

Supplemental Materials

Appendix B

How Much Should We Trust Instrumental Variable Estimates in Political Science? Practical Advice based on 67 Replicated Studies

28 October 2023

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Readme

- `est_ols` stores treatment effect estimates from the naive OLS estimation. ‘Analytic’ corresponds to analytic asymptotic standard errors (SEs) and confidence intervals (CIs). ‘Boot.c’ and ‘Boot.t’ represent inferential methods based on bootstrapped coefficients and bootstrapped t-statistics, respectively.
- `est_2sls` stores treatment effect estimates from the 2SLS estimation.
- `AR` stores results from the Anderson-Rubin test. The confidence region (CR) is produced by the inversion method. ‘AR.bounded = TRUE’ means that the CR is bounded and not empty.
- `F.stat` stores F statistics based on classic SEs (`F.standard`), H.W. robust SEs (`F.robust`), cluster-robust SEs (`F.cluster`), bootstrapped or cluster-bootstrapped SEs (`F.bootstrap`) and the effective F (`F.effective`). In the one-treatment-one-instrument case, `F.effective` is the same as `F.robust` (if there is no clustering structure) or `F.cluster` (if there is one).
- `rho` stores the partial correlation coefficient between the treatment and the predicted treatment from the first stage regression.
- `tf.cF` stores the results from the tF-cF procedure. Specifically, `cF` corresponds to the adjusted critical value based on the first stage (effective) F statistic for the subsequent t-test.
- `est_rf` stores the results from the reduced form regression. The control variables are partialled out.
- `est_fs` stores the results from the first stage regression. The control variables are partialled out.
- `p_iv` stores the number of instruments. `N` and `N_cl` stores the the number of observations and the number of clusters (if there is a clustering structure), respectively. `df` stores the degree of freedom from the 2SLS regression.
- `nvalues` stores the numbers of unique values in the outcome, treatment, and instrument.

APSR

Baccini and Weymouth (2021)

Replication Summary

Unit of analysis	county
Treatment	Manufacturing Layoffs
Instrument	Bartik instrument
Outcome	Change of Democratic Vote Share
Model	Table2(3)

```
df <- readRDS("./rawdata/apsr_baccini_etal_2021.rds")
D <- "msl_pc4y2"
Y <- "ddem_votes_pct1"
Z <- "bartik_leo5"
controls <- c("LAU_unemp_rate_4y", "pers_m_total_share_4y", "pers_coll_share_4y",
             "white_counties_4y", "msl_service_pc4y")
cl <- NULL
FE <- "id_state"
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

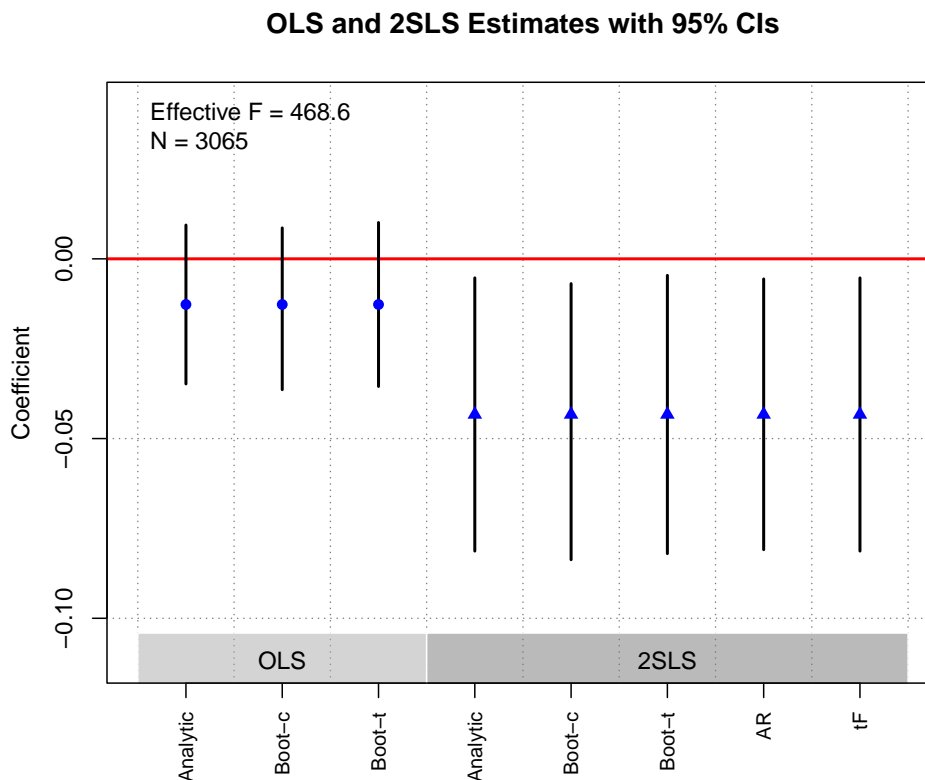
```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.0127 0.0113 -1.1240 -0.0348  0.0094  0.261
## Boot.c   -0.0127 0.0116 -1.0904 -0.0364  0.0086  0.262
## Boot.t   -0.0127 0.0113 -1.1240 -0.0355  0.0101  0.279
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.0433 0.0194 -2.2308 -0.0813 -0.0053  0.0257
## Boot.c   -0.0433 0.0192 -2.2535 -0.0837 -0.0069  0.0180
## Boot.t   -0.0433 0.0194 -2.2308 -0.0820 -0.0046  0.0230
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    5.0579    1.0000 3063.0000  0.0246
##
## $AR$ci.print
## [1] "[-0.0809, -0.0056]"
##
## $AR$ci
## [1] -0.0809 -0.0056
```

```

##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 1537.5647 468.6180 NA 492.8107 468.6180
##
## $rho
## [1] 0.5815
##
## $tF
## F cF Coef SE t CI2.5% CI97.5% p-value
## 468.6180 1.9600 -0.0433 0.0194 -2.2308 -0.0813 -0.0053 0.0257
##
## $est_rf
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## bartik_leo5 -4.5381 2.0355 0.0258 2.0041 -8.5967 -0.7369 0.018
##
## $est_fs
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## bartik_leo5 104.8786 4.8448 0 4.7244 95.7123 113.9969 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 3065
##
## $N_cl
## NULL
##
## $df
## [1] 3010
##
## $nvalues
## ddem_votes_pct1 msl_pc4y2 bartik_leo5
## [1,] 3062 2913 2771
##
## attr("class")
## [1] "ivDiag"

```

plot_coef(g)



Blattman et al. (2014)

Replication Summary

Unit of analysis	resident
Treatment	mass education campaign for dispute resolution
Instrument	assignment to treatment blocks
Outcome	serious land dispute
Model	Table9(8)

```
df <- readRDS("./rawdata/apsr_Blattman_etal_2014.rds")
df$district <- 0
for (i in 1:15) {df$district[which(df[,paste0("district",i)]==1)] <- i}
D <- "months_treated"
Y <- "fightweap_dummy"
Z <- c("block1", "block2", "block3")
controls <- c("ageover60", "age40_60", "age20_40",
             "yrs_edu", "female", "stranger", "christian",
             "minority", "cashearn_imputedhst", "noland",
             "land_sizehst", "farm_sizehst", "lndtake_dum",
             "housetake_dum", "vsmall", "small",
             "small2", "small3", "quartdummy", "cedulevel_bc",
```

```

"ctownhh_log_e1", "cwealthindex_bc", "cviol_experienced_bc",
"clndtake_bc", "cviol_scale_bc", "clandconf_scale_bc",
"cwitchcraft_scale_bc", "cpalaviol_imputed_bc",
"cprog_ldr_beliefs_bc", "cattitudes_tribe_bc",
"crelmarry_bc", "trainee")
cl <- "district"
FE <- "district"
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))

```

```

## $est_ols
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 7e-04 5e-04 1.2355 -4e-04  0.0018  0.2167
## Boot.c   7e-04 7e-04 0.9923 -9e-04  0.0018  0.3980
## Boot.t   7e-04 5e-04 1.2355 -7e-04  0.0020  0.2840
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 9e-04 5e-04 1.9157  0e+00  0.0018  0.0554
## Boot.c   9e-04 6e-04 1.4484 -6e-04  0.0019  0.2200
## Boot.t   9e-04 5e-04 1.9157 -2e-04  0.0020  0.0950
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
##  5.0886  3.0000 1896.0000  0.0016
##
## $AR$ci.print
## [1] "[0.0006, 0.0022]"
##
## $AR$ci
## [1] 0.0006 0.0022
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
##  2756.3845  2472.2847   234.3492   95.2132    52.1000
##
## $rho
## [1] 0.9039
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b

```

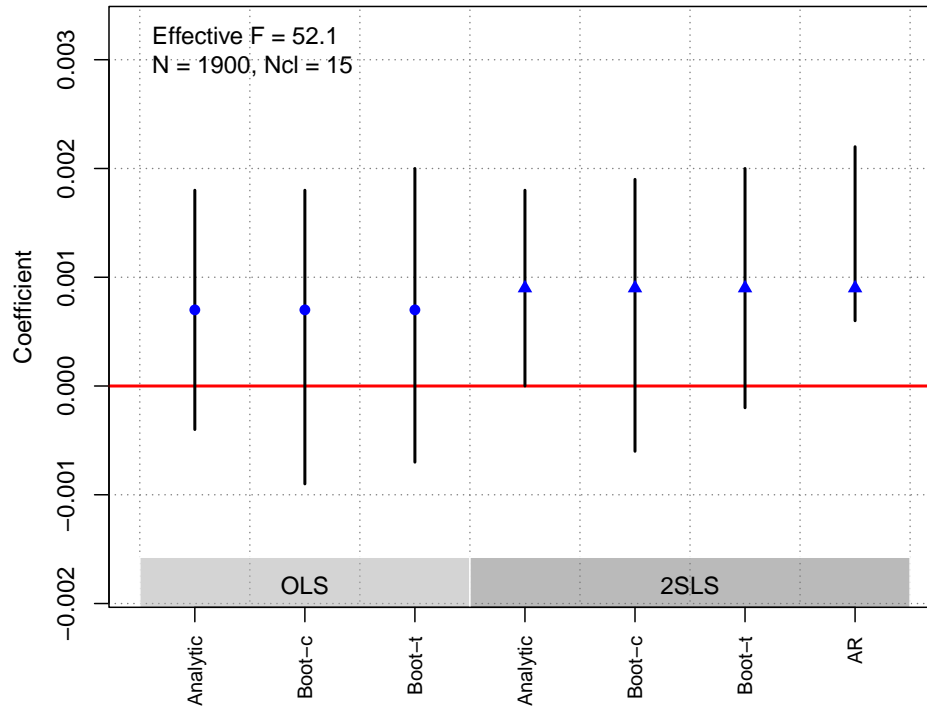
```

## block1 0.0263 0.0085 0.0020 0.0128 -0.0065 0.0448 0.090
## block2 0.0027 0.0099 0.7812 0.0136 -0.0256 0.0275 0.874
## block3 0.0085 0.0064 0.1816 0.0102 -0.0148 0.0256 0.344
##
## $est_fs
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## block1 20.0361 0.7567      0 1.2573 17.6152 22.5874 0.000
## block2 12.9786 1.7805      0 2.0999  8.9637 16.7281 0.000
## block3  6.7831 1.3081      0 1.8605  2.8911 10.5800 0.002
##
## $p_iv
## [1] 3
##
## $N
## [1] 1900
##
## $N_cl
## [1] 15
##
## $df
## [1] 14
##
## $nvalues
##      fightweap_dummy months_treated block1 block2 block3
## [1,]                2              34      2      2      2
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



Colantone and Stanig (2018)

Replication Summary

Unit of analysis	region
Treatment	regional-level import shock from China
Instrument	imports from China to the United States * local industrial structure
Outcome	leave share
Model	Table1(6)

```
df<-readRDS("./rawdata/apsr_Colantone_etal_2018.rds")
D <- 'import_shock'
Y <- "leave_share"
Z <- "instrument_for_shock"
controls <- c("immigrant_share", "immigrant_arrivals")
cl <- "fix"
FE <- "nuts1"
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 12.0854 3.8903 3.1066 4.4605 19.7104 0.0019
## Boot.c   12.0854 4.4192 2.7348 3.7385 21.4794 0.0020
```

```

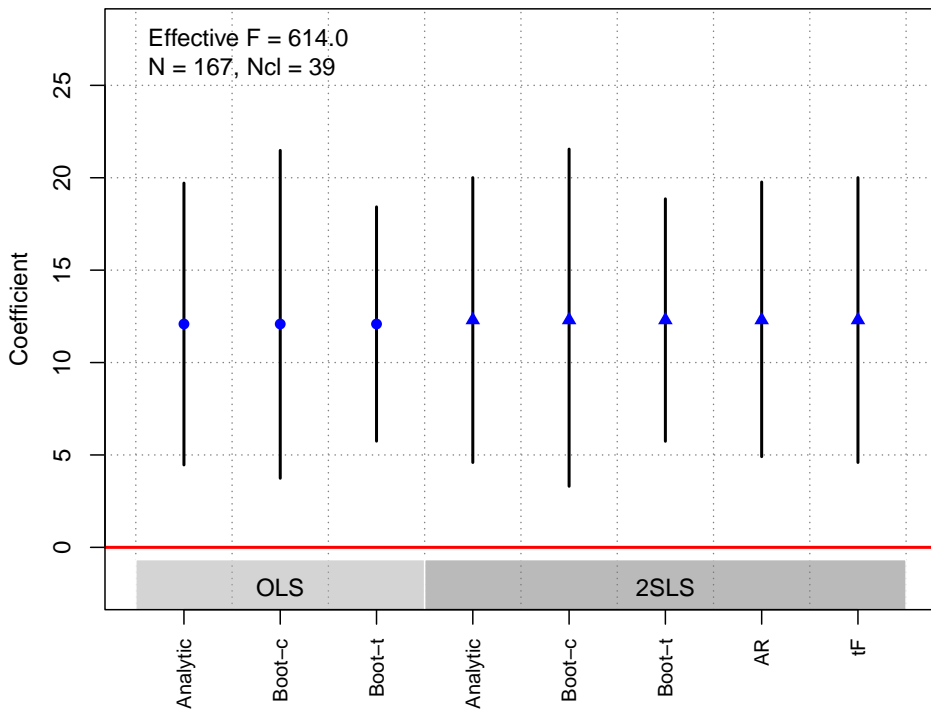
## Boot.t 12.0854 3.8903 3.1066 5.7465 18.4243 0.0000
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 12.2993 3.9320 3.1280 4.5926 20.0060 0.0018
## Boot.c   12.2993 4.6142 2.6655 3.3052 21.5500 0.0000
## Boot.t   12.2993 3.9320 3.1280 5.7402 18.8584 0.0000
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
## 10.5300 1.0000 165.0000 0.0014
##
## $AR$ci.print
## [1] "[4.9072, 19.7701]"
##
## $AR$ci
## [1] 4.9072 19.7701
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 2158.0662 792.4682 613.9804 581.1155 613.9804
##
## $rho
## [1] 0.9663
##
## $tF
##          F      cF      Coef      SE      t CI2.5% CI97.5% p-value
## 613.9804 1.9600 12.2993 3.9320 3.1280 4.5926 20.0060 0.0018
##
## $est_rf
##          Coef      SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## instrument_for_shock 1.5671 0.5015 0.0018 0.5904 0.4106 2.7361 0
##
## $est_fs
##          Coef      SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## instrument_for_shock 0.1274 0.0051 0 0.0053 0.1175 0.1385 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 167

```

```
##
## $N_cl
## [1] 39
##
## $df
## [1] 153
##
## $nvalues
##      leave_share import_shock instrument_for_shock
## [1,]          167           148                148
##
## attr("class")
## [1] "ivDiag"
```

plot_coef(g)

OLS and 2SLS Estimates with 95% CIs



Croke et al. (2016)

Replication Summary

Unit of analysis	individual
Treatment	education attainment
Instrument	access to the secondary education
Outcome	political participation
Model	Table2(b1)

```
df <-readRDS("./rawdata/apsr_Croke_etal_2016.rds")
D <- "edu"
Y <- "part_scale"
Z <- "treatment"
controls <-NULL
cl<- "district"
FE<- "year_survey"
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.0204 0.0078 -2.6133 -0.0357 -0.0051 0.009
## Boot.c   -0.0204 0.0072 -2.8262 -0.0318 -0.0048 0.014
## Boot.t   -0.0204 0.0078 -2.6133 -0.0357 -0.0051 0.012
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.098 0.0268 -3.6620 -0.1505 -0.0456 3e-04
## Boot.c   -0.098 0.0275 -3.5603 -0.1505 -0.0464 2e-03
## Boot.t   -0.098 0.0268 -3.6620 -0.1401 -0.0559 1e-03
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
## 16.1473 1.0000 1840.0000 0.0001
##
## $AR$ci.print
## [1] "[-0.1574, -0.0493]"
##
## $AR$ci
## [1] -0.1574 -0.0493
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 79.7552 78.2588 71.1356 69.0549 71.1356
##
## $rho
## [1] 0.2041
##
## $tF
##           F      cF      Coef      SE      t CI2.5% CI97.5% p-value
```

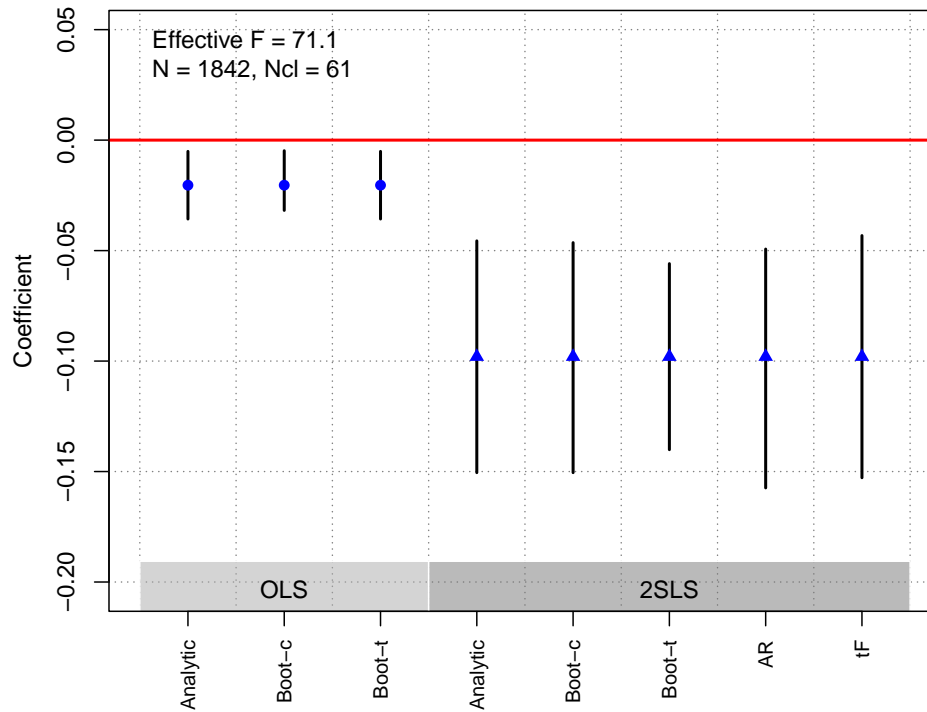
```

## 71.1356  2.0466 -0.0980  0.0268 -3.6620 -0.1528 -0.0432  0.0005
##
## $est_rf
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## treatment -0.0657 0.0164 1e-04 0.0165 -0.0969 -0.0323 0.002
##
## $est_fs
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## treatment 0.6708 0.0795 0 0.0807 0.5216 0.8418 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 1842
##
## $N_c1
## [1] 61
##
## $df
## [1] 1835
##
## $nvalues
##      part_scale edu treatment
## [1,]          7  7          5
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



Dower et al. (2018) (a)

Replication Summary

Unit of analysis	district*year
Treatment	frequency of unrest
Instrument	religious polarization
Outcome	peasant representation
Model	Table3(1)

```
df <- readRDS("./rawdata/aprs_Dower_etal_2018.rds")
D <- "afreq"
Y <- "peasantrepresentation_1864"
Z <- "religpolarf4_1870"
controls <- c("distance_moscow", "goodsoil", "lnurban", "lnpopn", "province_capital")
cl <- NULL
FE <- NULL
weights <- NULL
(g <- ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl, weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -3.8696 1.8013 -2.1483 -7.4001 -0.3391 0.0317
## Boot.c   -3.8696 1.8025 -2.1468 -7.4905 -0.5380 0.0240
```

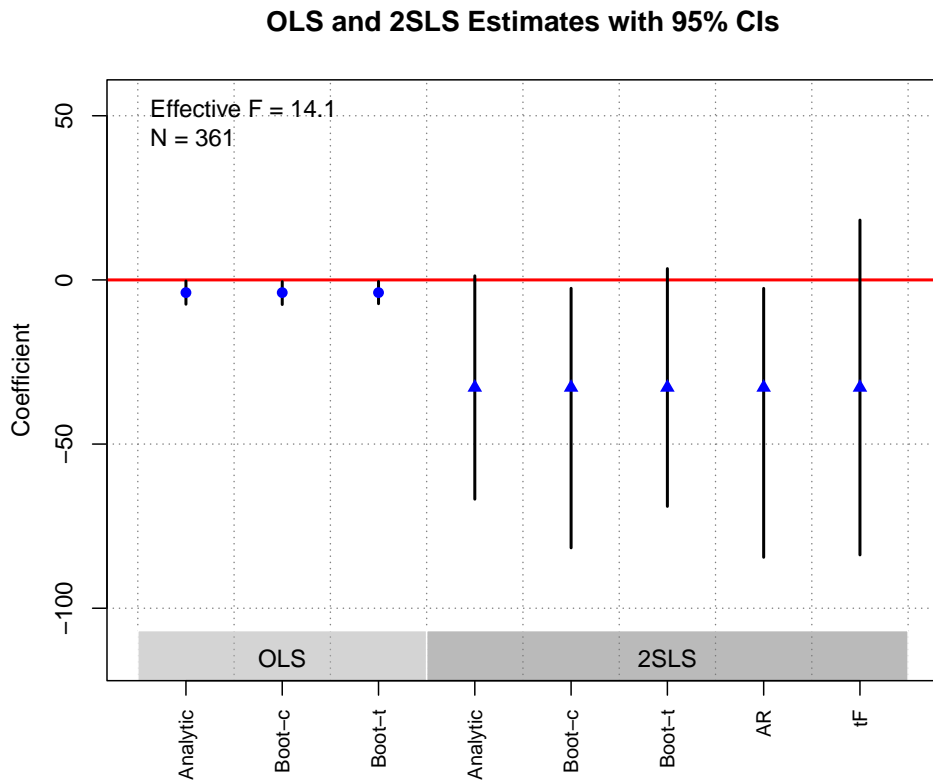
```

## Boot.t   -3.8696  1.8013 -2.1483 -7.2448  -0.4944  0.0280
##
## $est_2sls
##           Coef      SE      t  CI 2.5% CI 97.5% p.value
## Analytic -32.7701 17.3518 -1.8886 -66.7796  1.2393  0.0589
## Boot.c   -32.7701 19.3450 -1.6940 -81.6023 -2.5639  0.0240
## Boot.t   -32.7701 17.3518 -1.8886 -68.9771  3.4369  0.0620
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##  4.4669  1.0000 359.0000  0.0352
##
## $AR$ci.print
## [1] "[-84.4784, -2.5780]"
##
## $AR$ci
## [1] -84.4784 -2.5780
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
##  12.0237  14.0828      NA    14.8085    14.0828
##
## $rho
## [1] 0.1812
##
## $tF
##           F      cF      Coef      SE      t  CI2.5% CI97.5% p-value
##  14.0828  2.9384 -32.7701 17.3518 -1.8886 -83.7561 18.2159  0.2078
##
## $est_rf
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## religpolarf4_1870 -3.9279 1.8715  0.0358 1.8161  -7.569  -0.3608  0.024
##
## $est_fs
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## religpolarf4_1870 0.1199 0.0319  2e-04 0.0311  0.0611  0.1864  0
##
## $p_iv
## [1] 1
##
## $N
## [1] 361

```

```
##
## $N_cl
## NULL
##
## $df
## [1] 354
##
## $nvalues
##      peasantrepresentation_1864 afreq religpolarf4_1870
## [1,]                128      12                361
##
## attr("class")
## [1] "ivDiag"
```

plot_coef(g)



Dower et al. (2018) (b)

Replication Summary

Unit of analysis	district*year
Treatment	frequency of unrest
Instrument	religious polarization
Outcome	peasant representation
Model	Table1(2)

```
df <- readRDS("./rawdata/apsr_Dower_etal_2018.rds")
D <-"afreq"
Y <-"peasantrepresentation_1864"
Z <-"serfperc1"
controls <- c("distance_moscow", "goodsoil", "lnurban", "lnpopn", "province_capital")
cl <- NULL
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -4.2492 1.8297 -2.3224 -7.8353 -0.6631 0.0202
## Boot.c   -4.2492 1.8449 -2.3032 -8.0781 -0.9235 0.0200
## Boot.t   -4.2492 1.8297 -2.3224 -7.7849 -0.7135 0.0220
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -42.4545 8.4195 -5.0424 -58.9567 -25.9522      0
## Boot.c   -42.4545 9.2641 -4.5827 -64.3361 -28.3632      0
## Boot.t   -42.4545 8.4195 -5.0424 -59.4214 -25.4875      0
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
## 52.2466 1.0000 363.0000 0.0000
##
## $AR$ci.print
## [1] "[-63.3348, -28.4781]"
##
## $AR$ci
## [1] -63.3348 -28.4781
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 47.6256 51.0176 NA 50.2046 51.0176
##
## $rho
## [1] 0.3427
##
## $tF
##           F      cF      Coef      SE      t CI2.5% CI97.5% p-value
```

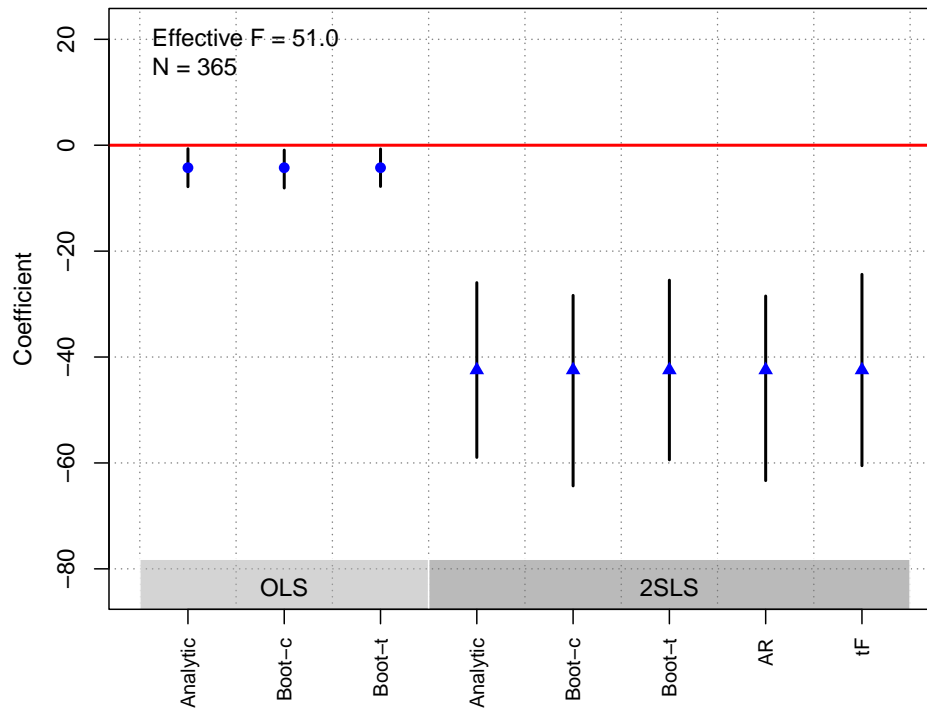
```

## 51.0176 2.1457 -42.4545 8.4195 -5.0424 -60.5204 -24.3885 0.0000
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## serfperc1 -11.7823 1.6414      0 1.7047 -15.0359 -8.2923      0
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## serfperc1  0.2775 0.0389      0 0.0392  0.2037  0.3499      0
##
## $p_iv
## [1] 1
##
## $N
## [1] 365
##
## $N_c1
## NULL
##
## $df
## [1] 358
##
## $nvalues
##      peasantrepresentation_1864 afreq serfperc1
## [1,]                          128  12      361
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



Gerber et al. (2010)

Replication Summary

Unit of analysis	individual
Treatment	aligning party identification with latent partisanship
Instrument	being sent mail
Outcome	voting and party alignment scale
Model	Table4(1)

```
df <- readRDS("./rawdata/apsr_Gerber_etal_2010.rds")
D <- "pt_id_with_lean"
Y <- "pt_voteevalalignindex"
Z <- "treat"
controls <- c("pre_lean_dem", "age", "age2", "regyear",
             "regyearmissing", "twonames", "combined_female",
             "voted2006", "voted2004", "voted2002", "voted2000",
             "voted1998", "voted1996", "interest", "pre_aligned_vh",
             "pre_direct_unemp", "pre_direct_econ", "pre_direct_bushap",
             "pre_direct_congapp")
cl <- NULL
FE <- NULL
weights <- NULL
(g <- ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
             cl =cl, weights=weights, cores = cores))
```

```

## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.5658 0.1709 3.3105 0.2308 0.9008 9e-04
## Boot.c   0.5658 0.1673 3.3821 0.2495 0.9045 0e+00
## Boot.t   0.5658 0.1709 3.3105 0.2353 0.8963 0e+00
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 3.8231 2.6392 1.4486 -1.3497 8.9960 0.1475
## Boot.c   3.8231 16.4557 0.2323 -6.4832 20.6934 0.1200
## Boot.t   3.8231 2.6392 1.4486 -1.8404 9.4867 0.1330
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
## 3.8593 1.0000 409.0000 0.0501
##
## $AR$ci.print
## [1] "[0.0227, Inf)"
##
## $AR$ci
## [1] 0.0227 Inf
##
## $AR$bounded
## [1] FALSE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 2.9926 3.1563 NA 3.2845 3.1563
##
## $rho
## [1] 0.0873
##
## $tF
##           F      cF      Coef      SE      t CI2.5% CI97.5% p-value
## 3.1563 18.6600 3.8231 2.6392 1.4486 -45.4249 53.0712 0.8791
##
## $est_rf
##           Coef      SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## treat 0.2742 0.1429 0.0551 0.1411 -0.0031 0.5497 0.056
##
## $est_fs
##           Coef      SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## treat 0.0717 0.0404 0.0756 0.0396 -0.0033 0.1515 0.07
##
## $p_iv

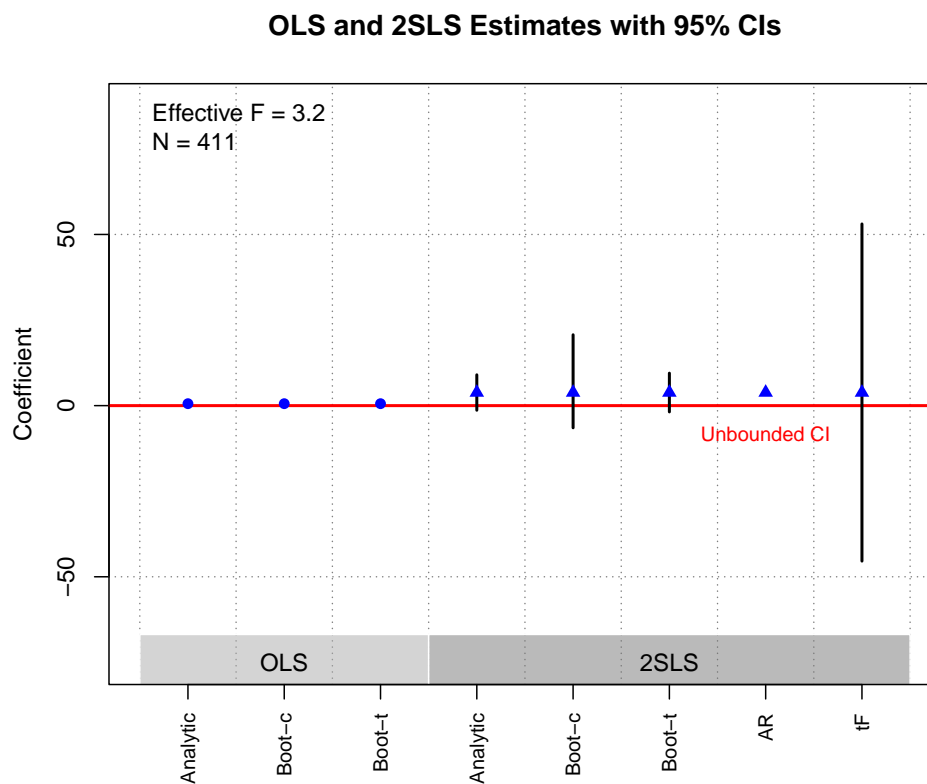
```

```

## [1] 1
##
## $N
## [1] 411
##
## $N_cl
## NULL
##
## $df
## [1] 390
##
## $nvalues
##      pt_voteevalalignindex pt_id_with_lean treat
## [1,]                10          2      2
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`



Hager et al. (2019)

Replication Summary

Unit of analysis	individual
Treatment	ethnic riots (destruction)
Instrument	distance to the nearest location where armored military vehicles were stolen
Outcome	prosocial behavior
Model	Figure6

```
df <- readRDS("./rawdata/apsr_Hager_etal_2019.rds")
D <-"affected"
Y <- "pd_in_scale"
Z <- "apc_min_distance"
controls <- NULL
cl <- NULL
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.2335 0.0675 -3.4582 -0.3658 -0.1011 5e-04
## Boot.c   -0.2335 0.0707 -3.3012 -0.3731 -0.0905 4e-03
## Boot.t   -0.2335 0.0675 -3.4582 -0.3756 -0.0914 4e-03
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.52 0.1416 -3.6733 -0.7975 -0.2425 2e-04
## Boot.c   -0.52 0.1468 -3.5419 -0.7947 -0.2255 0e+00
## Boot.t   -0.52 0.1416 -3.6733 -0.8060 -0.2341 0e+00
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
## 13.7909  1.0000 876.0000  0.0002
##
## $AR$ci.print
## [1] "[-0.8003, -0.2454]"
##
## $AR$ci
## [1] -0.8003 -0.2454
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
```

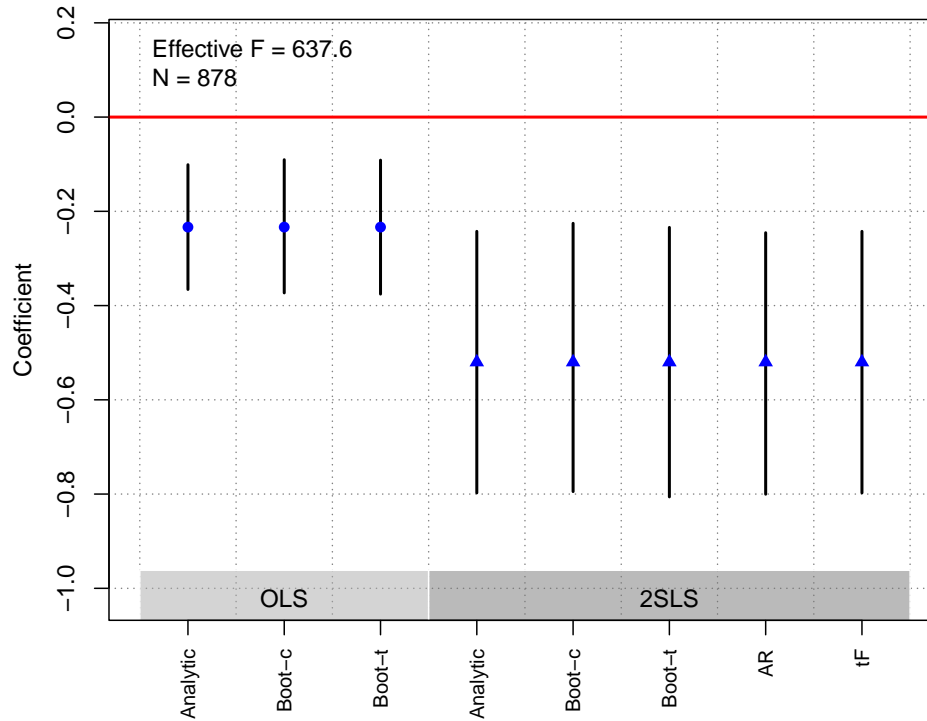
```

## F.standard F.robust F.cluster F.bootstrap F.effective
## 271.8565 637.5699 NA 644.1748 637.5699
##
## $rho
## [1] 0.4867
##
## $tF
## F cF Coef SE t CI2.5% CI97.5% p-value
## 637.5699 1.9600 -0.5200 0.1416 -3.6733 -0.7975 -0.2425 0.0002
##
## $est_rf
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## apc_min_distance 0.1011 0.0272 2e-04 0.0283 0.0445 0.1533 0
##
## $est_fs
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## apc_min_distance -0.1943 0.0077 0 0.0077 -0.2095 -0.1801 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 878
##
## $N_c1
## NULL
##
## $df
## [1] 876
##
## $nvalues
## pd_in_scale affected apc_min_distance
## [1,] 2 2 193
##
## attr("class")
## [1] "ivDiag"

```

plot_coef(g)

OLS and 2SLS Estimates with 95% CIs



Hager and Krakowski (2022)

Replication Summary

Unit of analysis	individual
Treatment	number of secret police officers
Instrument	number of corrupted Catholic priests
Outcome	resistance
Model	Table3(2)

```
df <- readRDS("./rawdata/apsr_Hager_Krakowski_2022.rds")

D <- "commanders"
Y <- "y"
Z <- "priests_continuous"
controls <- NULL
cl <- NULL
FE <- NULL
weights <- NULL
(g <- ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl, weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.1494 0.0751 1.9891 0.0022 0.2965 0.0467
```

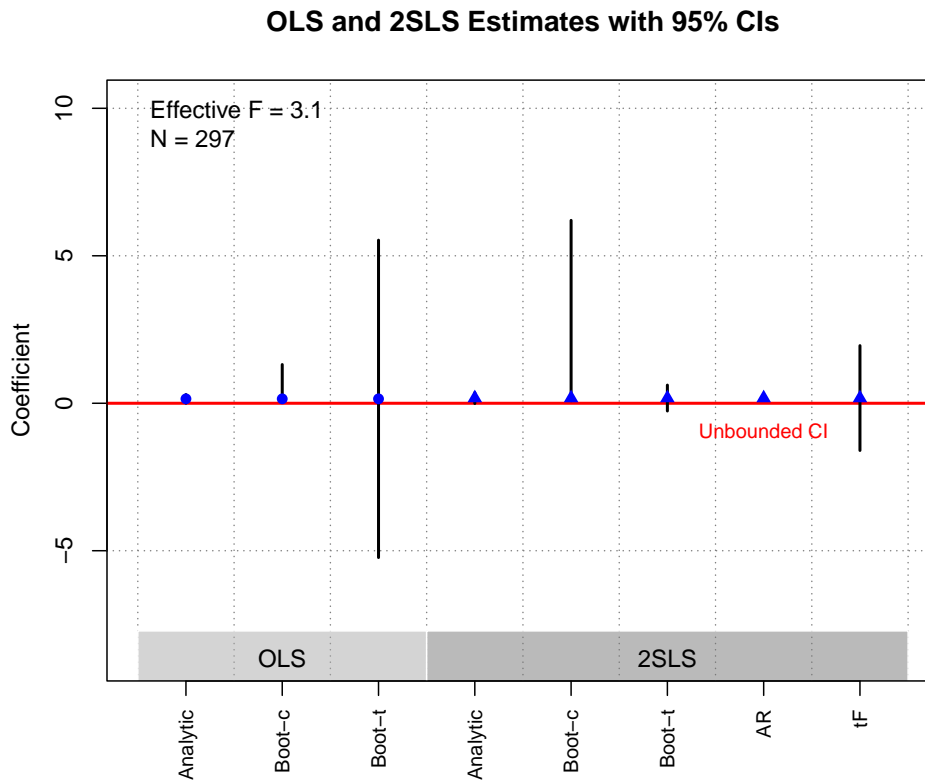
```

## Boot.c  0.1494 0.2844 0.5252 0.0594 1.3126 0.0000
## Boot.t  0.1494 0.0751 1.9891 -5.2312 5.5299 0.5040
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.1765 0.0952 1.8537 -0.0101 0.3632 0.0638
## Boot.c   0.1765 5.1233 0.0345 0.0837 6.2001 0.0060
## Boot.t   0.1765 0.0952 1.8537 -0.2629 0.6160 0.3570
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
##  8.7245  1.0000 295.0000 0.0034
##
## $AR$ci.print
## [1] "[0.0642, Inf)"
##
## $AR$ci
## [1] 0.0642  Inf
##
## $AR$bounded
## [1] FALSE
##
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
##  109.0543    3.1403         NA      3.5112      3.1403
##
## $rho
## [1] 0.5195
##
## $tF
##          F      cF      Coef      SE      t  CI2.5% CI97.5% p-value
##  3.1403 18.6600 0.1765 0.0952 1.8537 -1.6005 1.9535 0.8456
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## priests_continuous 0.4736 0.1603 0.0031 0.1733 0.1791 0.8722 0
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## priests_continuous 2.6827 1.5139 0.0764 1.4317 0.0273 5.1169 0.006
##
## $p_iv
## [1] 1
##
## $N

```

```
## [1] 297
##
## $N_c1
## NULL
##
## $df
## [1] 295
##
## $nvalues
##      y commanders priests_continuous
## [1,] 14          12                7
##
## attr("class")
## [1] "ivDiag"
```

`plot_coef(g)`



Kapoor and Magesan (2018)

Replication Summary

Unit of analysis	constituency*election
Treatment	number of independent candidates
Instrument	changes in entry costs
Outcome	voter turnout

Replication Summary

Model Table4(b5)

```
df<-readRDS("./rawdata/apsr_Kapoor_etal_2018.rds")
D <- 'CitCand'
Y <- "Turnout"
Z <- "UnScheduledDepChange"
controls <- c("CitCandBaseTrend", "CitCandBaseTrendSq", "CitCandBaseTrendCu",
             "CitCandBaseTrendQu", "TurnoutBaseTrend", "TurnoutBaseTrendSq",
             "TurnoutBaseTrendCu", "TurnoutBaseTrendQu", "LnElectors",
             "LagWinDist", "LagWinDistSq", "LagWinDistCu",
             "LagWinDistQu", "LagTightElection")
cl<- "constituency"
FE <- c("year","constituency")
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.0256 0.0110 -2.3216 -0.0472 -0.0040 0.0203
## Boot.c   -0.0256 0.0215 -1.1899 -0.0945 -0.0125 0.0000
## Boot.t   -0.0256 0.0110 -2.3216 -0.0554  0.0042 0.0750
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic  0.4864 0.2256  2.1562  0.0443  0.9285 0.0311
## Boot.c    0.4864 0.2648  1.8367  0.1224  1.1392 0.0020
## Boot.t    0.4864 0.2256  2.1562  0.1794  0.7934 0.0080
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    7.7339    1.0000 4295.0000  0.0054
##
## $AR$ci.print
## [1] "[0.1300, 1.1631]"
##
## $AR$ci
## [1] 0.1300 1.1631
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
```

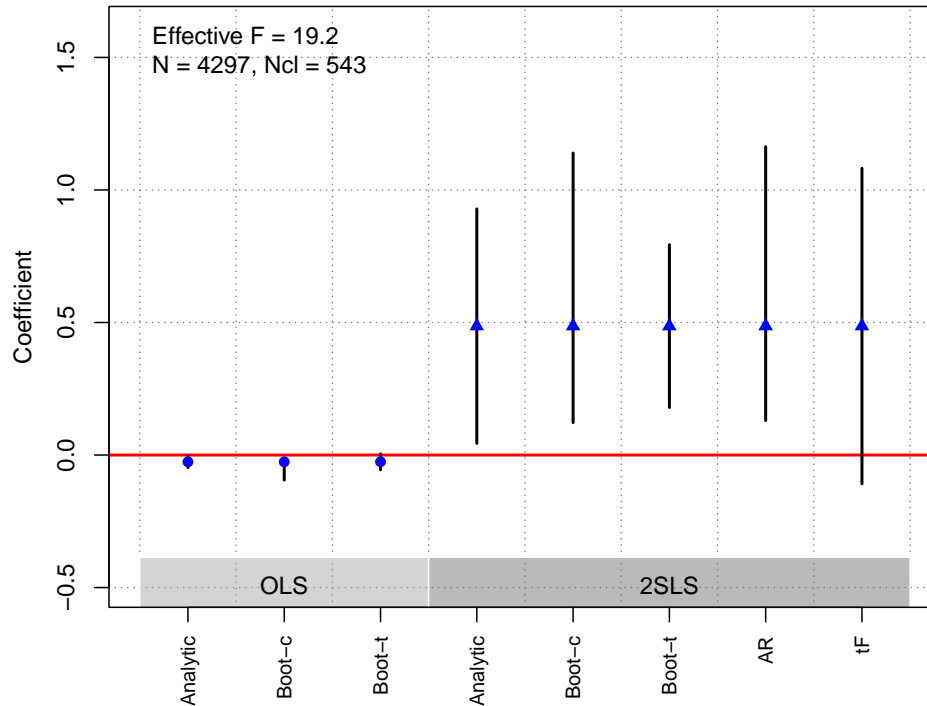
```

## F.standard F.robust F.cluster F.bootstrap F.effective
## 11.2301 23.7168 19.1635 19.7933 19.1635
##
## $rho
## [1] 0.0548
##
## $tF
## F cF Coef SE t CI2.5% CI97.5% p-value
## 19.1635 2.6390 0.4864 0.2256 2.1562 -0.1089 1.0817 0.1093
##
## $est_rf
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## UnScheduledDepChange -1.277 0.46 0.0055 0.4642 -2.1992 -0.3499 0.002
##
## $est_fs
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## UnScheduledDepChange -2.6256 0.5998 0 0.5902 -3.9271 -1.5206 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 4297
##
## $N_c1
## [1] 543
##
## $df
## [1] 542
##
## $nvalues
## Turnout CitCand UnScheduledDepChange
## [1,] 4293 68 2
##
## attr("class")
## [1] "ivDiag"

```

plot_coef(g)

OLS and 2SLS Estimates with 95% CIs



Kuipers and Sahn (2022)

Replication Summary

Unit of analysis	municipality* year
Treatment	civil service reform
Instrument	statewide assignment mandate
Outcome	descriptive representation on an unrestricted sample
Model	Table1(2)

```
df <- readRDS("./rawdata/aprs_kuipers_2022.rds")
df<-df%>%filter(occ=='blue_collar' & name=='white_x_native_born')
D <-"treat_actual"
Y <- "govt"
Z <- "treat_assign"
controls <-"pop"
cl <- NULL
FE <- c("YEAR","city")
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.0319 0.0156 -2.0467 -0.0625 -0.0014 0.0407
```

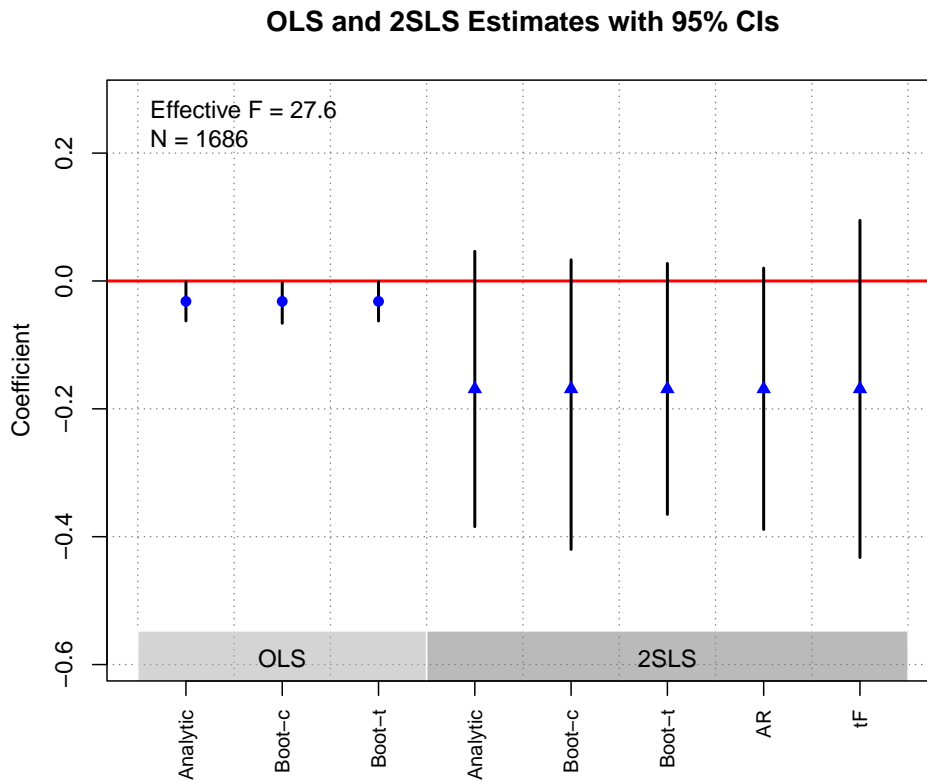
```

## Boot.c  -0.0319 0.0167 -1.9163 -0.0661 -0.0032 0.0280
## Boot.t  -0.0319 0.0156 -2.0467 -0.0626 -0.0012 0.0390
##
## $est_2sls
##          Coef      SE          t CI 2.5% CI 97.5% p.value
## Analytic -0.1689 0.1099 -1.5373 -0.3842  0.0464 0.1242
## Boot.c   -0.1689 0.1179 -1.4327 -0.4199  0.0331 0.0840
## Boot.t   -0.1689 0.1099 -1.5373 -0.3651  0.0274 0.0870
##
## $AR
## $AR$Fstat
##          F          df1          df2          p
##    3.0769      1.0000 1684.0000      0.0796
##
## $AR$ci.print
## [1] "[-0.3886, 0.0201]"
##
## $AR$ci
## [1] -0.3886  0.0201
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
##    32.4157    27.5670         NA    23.9060    27.5670
##
## $rho
## [1] 0.153
##
## $tF
##          F          cF      Coef      SE          t CI2.5% CI97.5% p-value
##    27.5670    2.3999 -0.1689  0.1099 -1.5373 -0.4326  0.0948  0.2093
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## treat_assign -0.0254 0.0162  0.116 0.017 -0.0611  0.0052  0.084
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## treat_assign 0.1504 0.0286      0 0.0308  0.0969  0.2157      0
##
## $p_iv
## [1] 1
##
## $N

```

```
## [1] 1686
##
## $N_cl
## NULL
##
## $df
## [1] 1352
##
## $nvalues
##      govt treat_actual treat_assign
## [1,] 658           2           2
##
## attr("class")
## [1] "ivDiag"
```

plot_coef(g)



Laitin and Ramachandran (2016)

Replication Summary

Unit of analysis	country
Treatment	language choice
Instrument	geographic distance from the origins of writing
Outcome	human development index

Replication Summary

Model Table10(10)

```
df <-readRDS("./rawdata/apsr_Laitin_2016.rds")
D <-"avgdistance_delta50"
Y <- "zhdi_2010"
Z <- "DIST_BGNC"
controls <- c("cdf2003","ln_GDP_Indp", "edes1975",
             "America","xconst")
cl<- NULL
FE<- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -1.3676 0.1884 -7.2594 -1.7369 -0.9984      0
## Boot.c   -1.3676 0.1957 -6.9878 -1.7619 -0.9725      0
## Boot.t   -1.3676 0.1884 -7.2594 -1.7818 -0.9535      0
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -1.3815 0.2963 -4.6618 -1.9623 -0.8007      0
## Boot.c   -1.3815 0.3029 -4.5608 -1.9614 -0.7515      0
## Boot.t   -1.3815 0.2963 -4.6618 -1.9571 -0.8059      0
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
## 11.4476  1.0000 135.0000 0.0009
##
## $AR$ci.print
## [1] "[-1.9505, -0.7295]"
##
## $AR$ci
## [1] -1.9505 -0.7295
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
## 55.1871     32.4040      NA         33.5596     32.4040
##
```

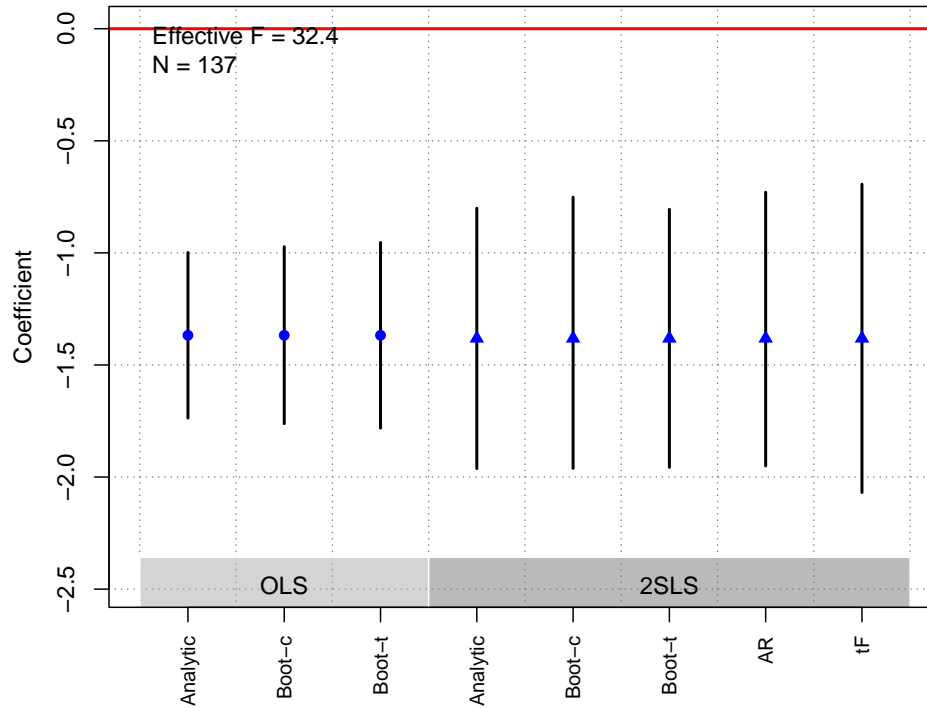
```

## $rho
## [1] 0.5459
##
## $tF
##      F      cF    Coef      SE      t  CI2.5% CI97.5% p-value
## 32.4040  2.3208 -1.3815  0.2963 -4.6618 -2.0692 -0.6938  0.0001
##
## $est_rf
##           Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## DIST_BGNC -1e-04 0 9e-04 0 -2e-04 0 0
##
## $est_fs
##           Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## DIST_BGNC 1e-04 0 0 0 1e-04 1e-04 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 137
##
## $N_cl
## NULL
##
## $df
## [1] 130
##
## $nvalues
##      zhdi_2010 avgdistance_delta50 DIST_BGNC
## [1,]      121          93      134
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



Meredith (2013)

Replication Summary

Unit of analysis	down-ballot race
Treatment	Democratic governor
Instrument	governor's home county
Outcome	down-ballot Democratic candidates' vote share
Model	Table3(5)

```
df <- readRDS("./rawdata/apsr_Meredith_2013.rds")
Y <- "DemShareDB_res"
D <- "DemShareGOV_res"
Z <- "HomeGOV_res"
controls <- "HomeDB_res"
cl <- "fips"
FE <- NULL
weights <- NULL
(g <- ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl, weights=weights, cores = cores))
```

```
## $est_ols
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.2634 0.0128 20.5976 0.2383  0.2884      0
## Boot.c   0.2634 0.0127 20.7067 0.2400  0.2910      0
```

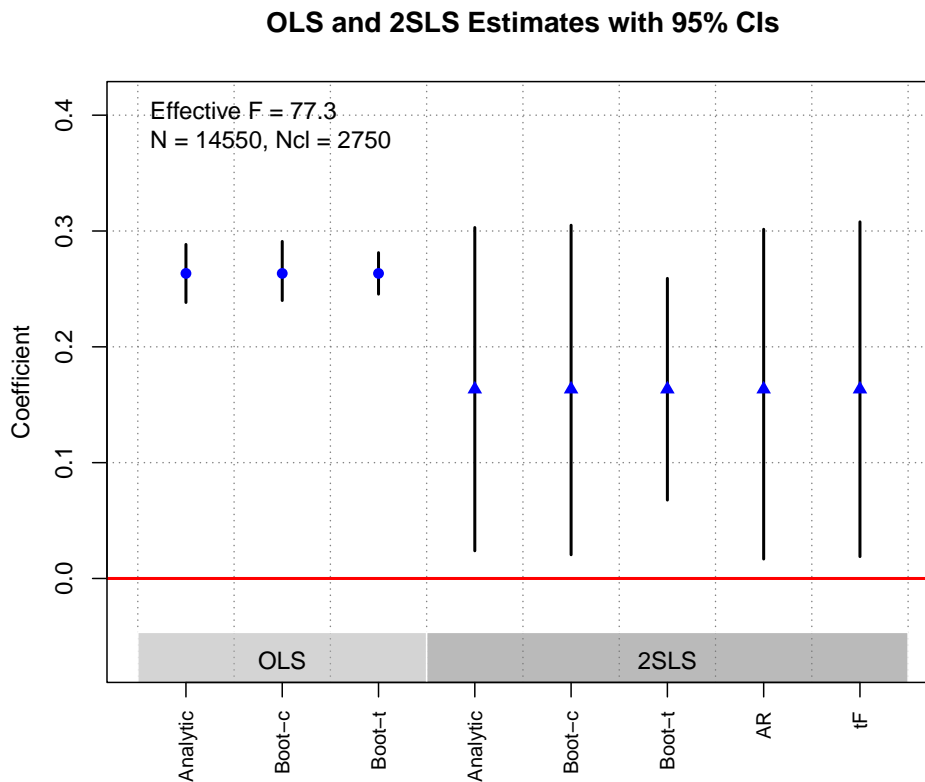
```

## Boot.t  0.2634 0.0128 20.5976  0.2455  0.2813      0
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.1634 0.0712 2.2959  0.0239  0.3030  0.0217
## Boot.c   0.1634 0.0749 2.1832  0.0205  0.3050  0.0300
## Boot.t   0.1634 0.0712 2.2959  0.0677  0.2591  0.0010
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
##    4.6123    1.0000 14548.0000    0.0318
##
## $AR$ci.print
## [1] "[0.0168, 0.3015]"
##
## $AR$ci
## [1] 0.0168 0.3015
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
##    284.9652    141.9189    77.2953    76.8134    77.2953
##
## $rho
## [1] 0.1386
##
## $tF
##          F      cF      Coef      SE      t CI2.5% CI97.5% p-value
##    77.2953  2.0300  0.1634  0.0712  2.2959  0.0189  0.3079  0.0266
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## HomeGOV_res 0.0062 0.0029  0.0317 0.0029  7e-04  0.0115  0.03
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## HomeGOV_res 0.0379 0.0043  0 0.0043  0.0298  0.0467  0
##
## $p_iv
## [1] 1
##
## $N
## [1] 14550

```

```
##
## $N_cl
## [1] 2750
##
## $df
## [1] 14547
##
## $nvalues
##      DemShareDB_res DemShareGOV_res HomeGOV_res
## [1,]      14550      14550      1466
##
## attr("class")
## [1] "ivDiag"
```

plot_coef(g)



Nellis and Siddiqui (2018)

Replication
Summary

Unit of analysis	district*election
Treatment	the proportion of MNA seats in a district won by secularist candidates
Instrument	narrow victory by secular parties in a district
Outcome	religious violence

Replication
Summary

Model Table2(1)

```
df<-readRDS("./rawdata/apsr_Nellis_etal_2018.rds")
D <- 'secular_win'
Y <- "any_violence"
Z <- "secular_close_win"
controls <- "secular_close_race"
cl <- "cluster_var"
FE <- "pro"
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))

## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.015 0.0364 -0.4107 -0.0863  0.0564  0.6813
## Boot.c   -0.015 0.0369 -0.4055 -0.0844  0.0601  0.6660
## Boot.t   -0.015 0.0364 -0.4107 -0.0700  0.0401  0.6070
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.6603 0.2154 -3.0658 -1.0825 -0.2382  0.0022
## Boot.c   -0.6603 0.2556 -2.5832 -1.0888 -0.0691  0.0260
## Boot.t   -0.6603 0.2154 -3.0658 -1.0190 -0.3016  0.0100
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
## 12.2950  1.0000 435.0000  0.0005
##
## $AR$ci.print
## [1] "[-1.1557, -0.2813]"
##
## $AR$ci
## [1] -1.1557 -0.2813
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
## 22.0208     60.0400     53.9103     41.7317     53.9103
##
```

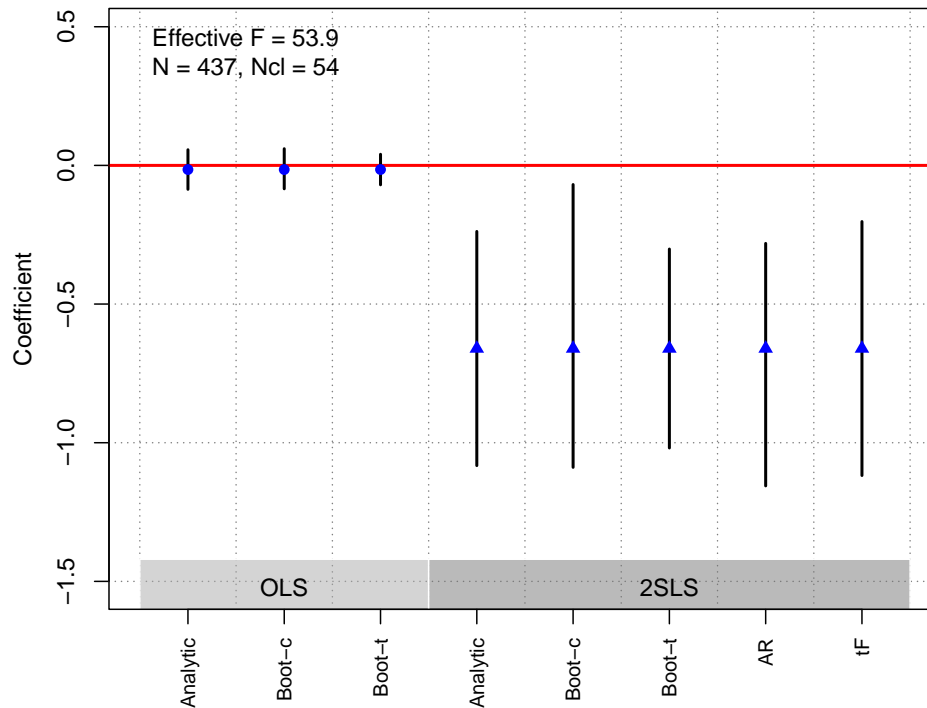
```

## $rho
## [1] 0.2207
##
## $tF
##      F      cF    Coef      SE      t  CI2.5% CI97.5% p-value
## 53.9103  2.1258 -0.6603  0.2154 -3.0658 -1.1182 -0.2025  0.0047
##
## $est_rf
##              Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## secular_close_win -0.5965 0.1711 5e-04 0.1998 -0.8619 -0.0804 0.026
##
## $est_fs
##              Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## secular_close_win 0.9034 0.123 0 0.1398 0.6222 1.1849 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 437
##
## $N_cl
## [1] 54
##
## $df
## [1] 430
##
## $nvalues
##      any_violence secular_win secular_close_win
## [1,]           2           26           17
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



Ritter and Conrad (2016)

Replication Summary	
Unit of analysis	province in 54 African countries*day
Treatment	mobilized dissent
Instrument	rainfall
Outcome	repression
Model	Table1(3b)

```
df <- readRDS("./rawdata/aprs_Ritter_etal_2016.rds")
D <- "dissentcount"
Y <- "represscount"
Z <- c("lograin", "rainannualpct")
controls <- "urban_mean"
cl <- NULL
FE <- NULL
weights <- NULL
(g <- ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl, weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.1885 0.0067 28.0525 0.1754 0.2017      0
## Boot.c   0.1885 0.0068 27.7793 0.1758 0.2016      0
```

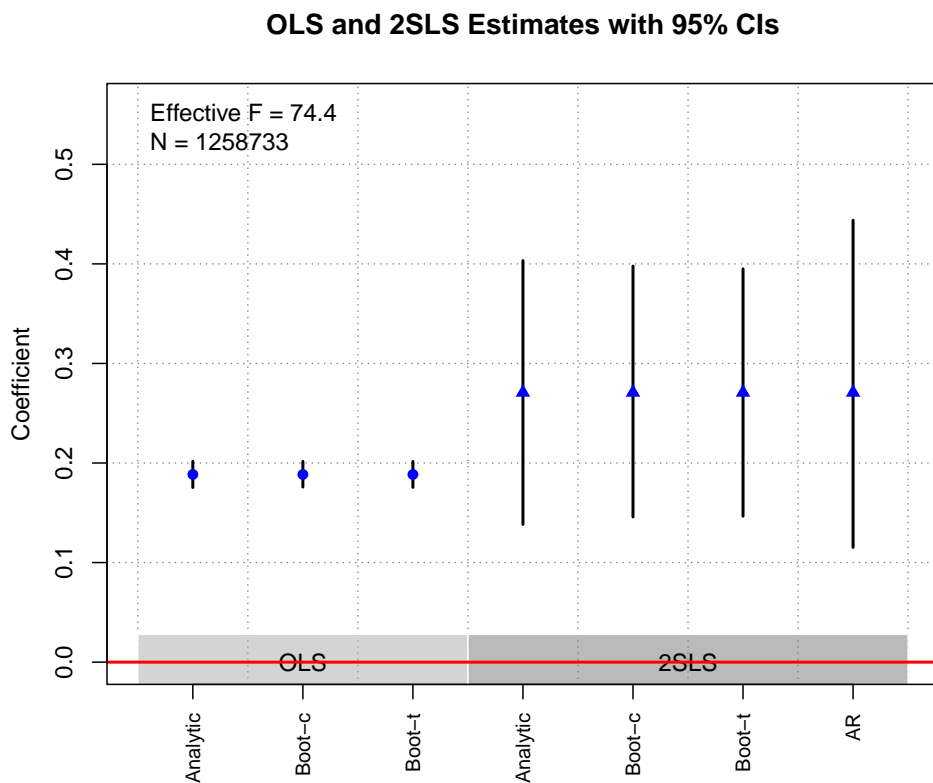
```

## Boot.t  0.1885 0.0067 28.0525  0.1755  0.2016      0
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.2708 0.0676 4.0058  0.1383  0.4033  1e-04
## Boot.c   0.2708 0.0673 4.0257  0.1458  0.3978  0e+00
## Boot.t   0.2708 0.0676 4.0058  0.1466  0.3950  0e+00
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
## 8.36210e+00 2.00000e+00 1.25873e+06 2.00000e-04
##
## $AR$ci.print
## [1] "[0.1153, 0.4438]"
##
## $AR$ci
## [1] 0.1153 0.4438
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 58.3505 73.6819 NA 72.6998 74.3587
##
## $rho
## [1] 0.0096
##
## $est_rf
##          Coef      SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## lograin 0.0001 0.0000 0.0000 0.0000 0.0001 0.0002 0.000
## rainannualpct -0.0092 0.0059 0.1199 0.0061 -0.0201 0.0028 0.122
##
## $est_fs
##          Coef      SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## lograin 0.0005 0.0000 0e+00 0.0000 0.0004 0.0006 0
## rainannualpct -0.0250 0.0065 1e-04 0.0069 -0.0382 -0.0116 0
##
## $p_iv
## [1] 2
##
## $N
## [1] 1258733
##
## $N_c1

```

```
## NULL
##
## $df
## [1] 1258730
##
## $nvalues
##      represscount dissentcount lograin rainannualpct
## [1,]           3           5 390194      593785
##
## attr(,"class")
## [1] "ivDiag"
```

plot_coef(g)



AJPS

Barth et al. (2015)

Replication Summary

Unit of analysis	country*year
Treatment	wage inequality
Instrument	adjusted bargaining coverage; effective number of union confederations

Replication Summary

Outcome welfare support
Model Table4(1)

```
df<- readRDS("./rawdata/ajps_Barth_2015.rds")
D <-"ld9d1"
Y <- "welfareleft"
Z <- c("l2ip_adjcov5", "l2ip_enucfs")
controls <- c("lgdpr", "lelderly", "llntexp", "lud", "ludsq",
              "lechp", "lnet", "lannual", "ltrend", "ltrendsq")
cl <- FE <- "countrynumber"
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.7755 0.2358 -3.2886 -1.2376 -0.3133 0.001
## Boot.c   -0.7755 0.3286 -2.3598 -1.4249 -0.1264 0.022
## Boot.t   -0.7755 0.2358 -3.2886 -1.2427 -0.3082 0.003
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -1.4265 0.7779 -1.8339 -2.9511 0.0981 0.0667
## Boot.c   -1.4265 2.0723 -0.6884 -4.0218 2.9224 0.3200
## Boot.t   -1.4265 0.7779 -1.8339 -3.0006 0.1477 0.0740
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
## 3.6053 2.0000 114.0000 0.0303
##
## $AR$ci.print
## [1] "[-4.0005, -0.1197]"
##
## $AR$ci
## [1] -4.0005 -0.1197
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 9.7741 15.0268 11.5754 3.0487 8.1611
##
```

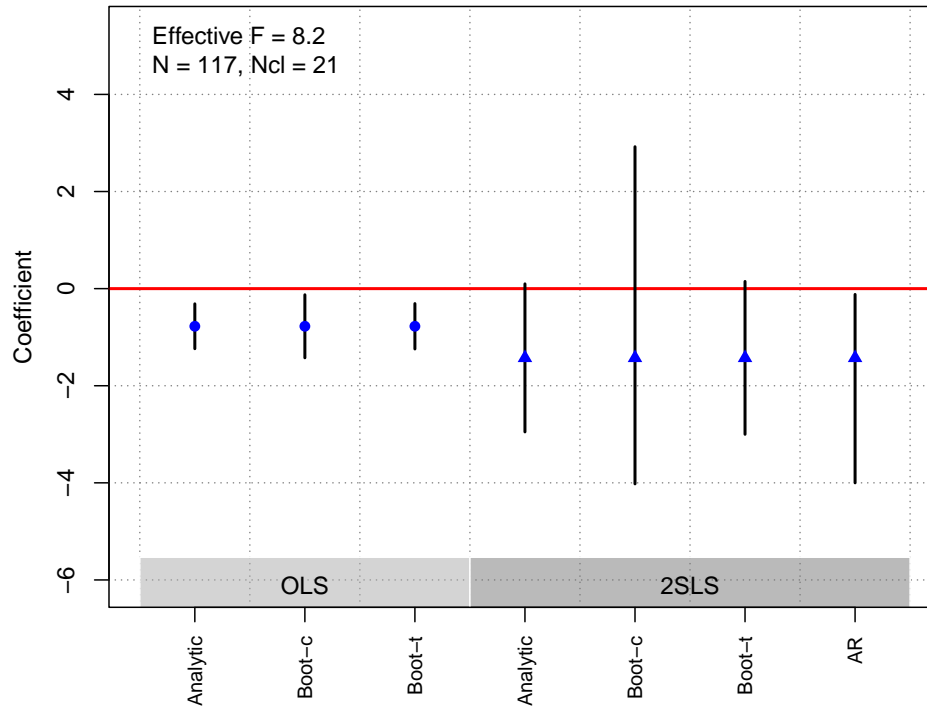
```

## $rho
## [1] 0.4345
##
## $est_rf
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## l2ip_adjcov5 0.0184 0.0124  0.1377 0.0193 -0.0262   0.0507   0.380
## l2ip_enucfs  0.1687 0.2420  0.4858 0.3867 -0.8098   0.7905   0.768
##
## $est_fs
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## l2ip_adjcov5 -0.0096 0.0046  0.0383 0.0070 -0.0267  -0.0002   0.044
## l2ip_enucfs  -0.1542 0.0777  0.0473 0.1028 -0.2885   0.1103   0.190
##
## $p_iv
## [1] 2
##
## $N
## [1] 117
##
## $N_c1
## [1] 21
##
## $df
## [1] 20
##
## $nvalues
##      welfareleft ld9d1 l2ip_adjcov5 l2ip_enucfs
## [1,]           117   117           106           112
##
## attr("class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



Blair et al. (2022)

Replication Summary

Unit of analysis	UN peacekeeping operations event level
Treatment	fragmentation of any given PKO mandate
Instrument	average fragmentation of all ongoing PKO mandates
Outcome	process performance
Model	TableD7(3)

```
df <- readRDS("./rawdata/ajps_Blair_2022.rds")
df <- as.data.frame(df)
D <- "L_avg"
Y <- "sh_perfassist_pb"
Z <- "L_fract_assistv3"
controls <- c("L_experman_assist_pbv3", "L_numtask_assist_pbv3", "L_lntot",
              "L_deployment", "L_lnpop", "L_lngdp", "L_ucdpconflictspell", "L_polity")
cl <- NULL
FE <- c("date3", "iso3n")
weights <- NULL
(g <- ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
             cl =cl, weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
```

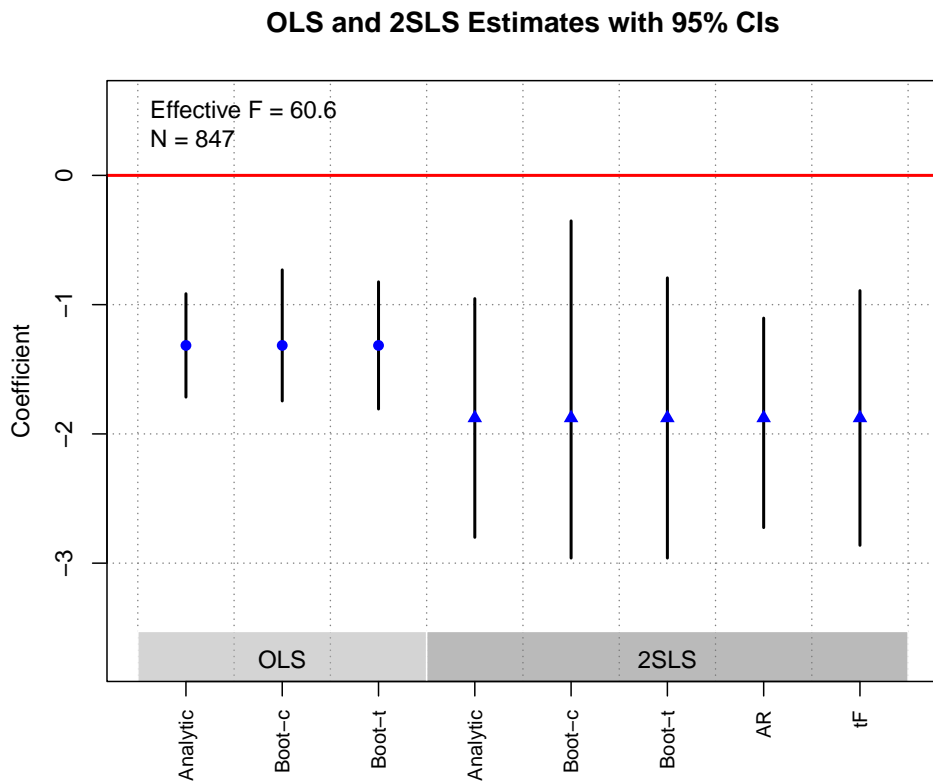
```

## Analytic -1.3155 0.2040 -6.4481 -1.7153 -0.9156 0
## Boot.c -1.3155 0.2527 -5.2056 -1.7451 -0.7310 0
## Boot.t -1.3155 0.2040 -6.4481 -1.8073 -0.8237 0
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -1.8768 0.4711 -3.9841 -2.8001 -0.9535 0.0001
## Boot.c   -1.8768 0.6586 -2.8498 -2.9605 -0.3521 0.0140
## Boot.t   -1.8768 0.4711 -3.9841 -2.9606 -0.7930 0.0010
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
## 20.4937  1.0000 845.0000  0.0000
##
## $AR$ci.print
## [1] "[-2.7247, -1.1042]"
##
## $AR$ci
## [1] -2.7247 -1.1042
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
## 186.0679    60.6442          NA    24.4332    60.6442
##
## $rho
## [1] 0.4793
##
## $tF
##          F      cF      Coef      SE      t CI2.5% CI97.5% p-value
## 60.6442  2.0913 -1.8768  0.4711 -3.9841 -2.8619 -0.8917  0.0002
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## L_fract_assistv3 1.805 0.464 1e-04 0.751  0.3215  3.4116  0.014
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## L_fract_assistv3 -0.9617 0.1235  0 0.1946 -1.5068  -0.71  0
##
## $p_iv
## [1] 1
##

```

```
## $N
## [1] 847
##
## $N_c1
## NULL
##
## $df
## [1] 624
##
## $nvalues
##      sh_perfassist_pb L_avg L_fract_assistv3
## [1,]           56    55           222
##
## attr("class")
## [1] "ivDiag"
```

`plot_coef(g)`



Carnegie and Marinov (2017)

Replication Summary

Unit of analysis	country*year
Treatment	foreign aid
Instrument	being a former colony of one of the Council members

Replication Summary

Outcome CIRI Human Empowerment index
Model Table1(2)

```
df<-readRDS("./rawdata/ajps_Carnegie_etal_2017.rds")
D <-"EV"
Y <- "new_empinxavg"
Z <- "l2CPcol2"
controls <- c( "covloggdpc", "covloggdpcF", "covloggdpcC",
               "covdemregionF", "covdemregion", "coviNY_GDP_PETR_RT_ZSF",
               "coviNY_GDP_PETR_RT_ZS", "covwvs_reIF", "covwvs_rel",
               "covwdi_imp", "covwdi_fdiF", "covwdi_fdi",
               "covwdi_expF", "covwdi_exp", "covihme_ayemF", "covihme_ayem")
cl<-c("year", "ccode")
FE <- c("year", "ccode")
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.1903 0.1376 1.3831 -0.0794  0.4601  0.1666
## Boot.c   0.1903 0.0761 2.5001  0.0457  0.3463  0.0040
## Boot.t   0.1903 0.1376 1.3831  0.0449  0.3358  0.0190
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 1.7054 0.8436 2.0217  0.0520  3.3589  0.0432
## Boot.c   1.7054 9.7806 0.1744 -5.7868  8.5463  0.1880
## Boot.t   1.7054 0.8436 2.0217  0.3318  3.0791  0.0240
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    2.7312    1.0000 1790.0000  0.0986
##
## $AR$ci.print
## [1] "[-0.5722, 4.0169]"
##
## $AR$ci
## [1] -0.5722  4.0169
##
## $AR$bounded
## [1] TRUE
##
##
```

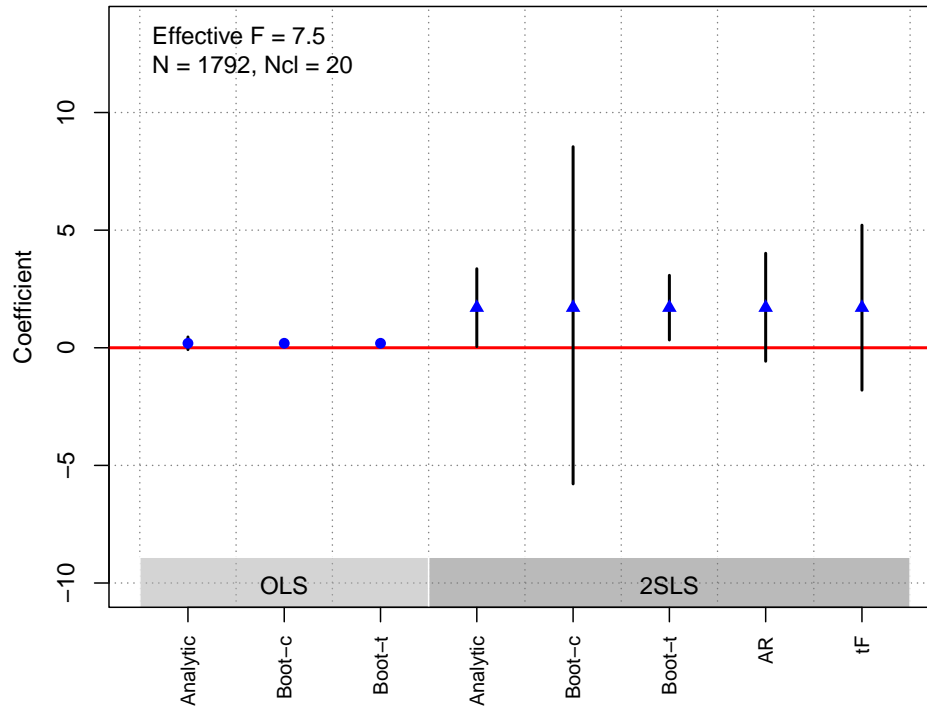
```

## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 4.5101 4.5766 7.5007 3.8839 7.5007
##
## $rho
## [1] 0.0523
##
## $tF
## F cF Coef SE t CI2.5% CI97.5% p-value
## 7.5007 4.1570 1.7054 0.8436 2.0217 -1.8014 5.2123 0.3405
##
## $est_rf
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## l2CPcol2 0.2632 0.16 0.0998 0.1934 -0.0659 0.6537 0.14
##
## $est_fs
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## l2CPcol2 0.1543 0.0564 0.0062 0.0783 -0.0163 0.2959 0.076
##
## $p_iv
## [1] 1
##
## $N
## [1] 1792
##
## $N_cl
## [1] 20
##
## $df
## [1] 19
##
## $nvalues
## new_empinxavg EV l2CPcol2
## [1,] 57 1601 2
##
## attr("class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



Chong et al. (2019)

Replication Summary

Unit of analysis	household
Treatment	actual proportion of households treated in the locality
Instrument	treatment assignment in get-out-to-vote campaigns
Outcome	voted in 2013 presidential election
Model	Table4(1)

```
df <- readRDS("../rawdata/ajps_Chong_etal_2019.rds")
D <- "ratio_treat"
Y <- "elecc_presid2013"
Z <- c("D2D30", "D2D40", "D2D50")
controls <- c("age", "married", "children", "num_children",
             "employed", "languag", "yrseeduc", "bornloc",
             "hh_asset_index", "log_pop", "mujeres_perc",
             "pob_0_14_perc", "pob_15_64_perc", "pob_65mas_perc",
             "analfabetos_perc", "asiste_escuela_perc",
             "TASA_women", "TASA_men", "electricidad_perc",
             "agua_perc", "desague_perc", "basura_perc",
             "fono_fijo_perc", "fono_cel_perc", "ocupantes",
             "Rural", "distancia2_final", "db_age",
             "db_married", "db_children", "db_num_children",
             "db_employed", "db_languag", "db_yrseeduc",
```

```

      "db_bornloc", "db_hh_asset_index", "db_log_pop",
      "db_mujeres_perc", "db_pob_0_14_perc",
      "db_pob_15_64_perc", "db_pob_65mas_perc",
      "db_analfabetos_perc", "db_asiste_escuela_perc",
      "db_TASA_women", "db_TASA_men", "db_electricidad_perc",
      "db_agua_perc", "db_desague_perc", "db_basura_perc",
      "db_fono_fijo_perc", "db_fono_cel_perc",
      "db_ocupantes", "db_Rural", "db_distancia2_final",
      "dpto1", "elecc_presid2008", "db_elecc_presid2008")
cl <- "loc"
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))

```

```

## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0715 0.0421 1.6984 -0.0110  0.1541  0.0894
## Boot.c   0.0715 0.0452 1.5826 -0.0226  0.1502  0.1220
## Boot.t   0.0715 0.0421 1.6984 -0.0011  0.1442  0.0520
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.1242 0.0527 2.3584  0.0210  0.2275  0.0184
## Boot.c   0.1242 0.0573 2.1680  0.0043  0.2339  0.0420
## Boot.t   0.1242 0.0527 2.3584  0.0390  0.2094  0.0040
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    2.5349    3.0000 3346.0000  0.0551
##
## $AR$ci.print
## [1] "[-0.0022, 0.2791]"
##
## $AR$ci
## [1] -0.0022  0.2791
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
##    1163.8658    270.5690    37.7653    31.5181    32.5611
##
## $rho

```

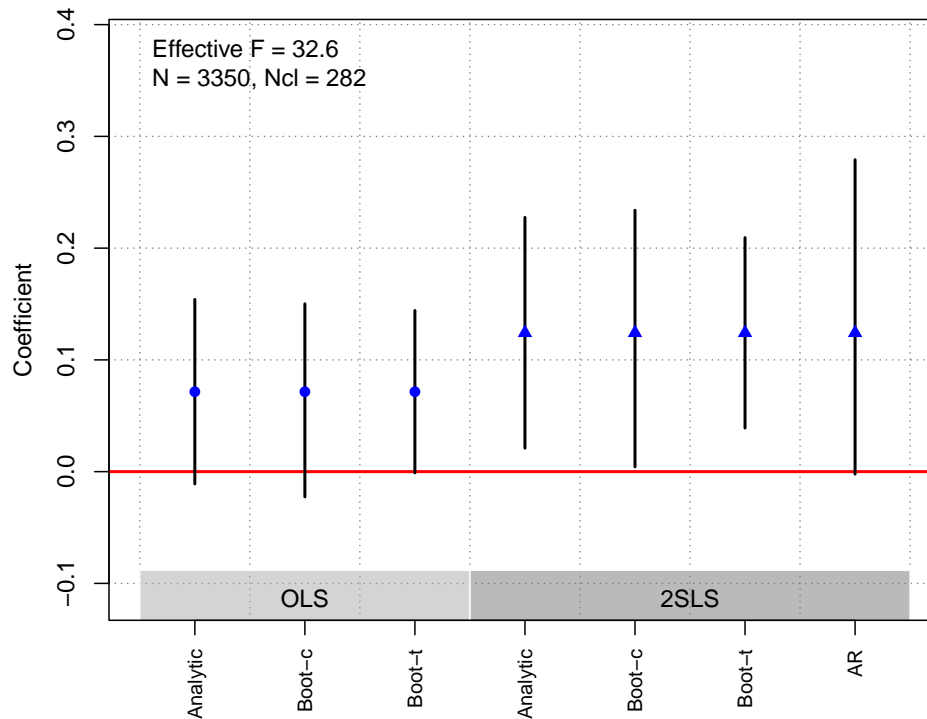
```

## [1] 0.7163
##
## $est_rf
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## D2D30 0.0194 0.0333  0.5611 0.0358 -0.0568  0.0813  0.610
## D2D40 0.0651 0.0243  0.0075 0.0276  0.0054  0.1184  0.036
## D2D50 0.0190 0.0277  0.4940 0.0297 -0.0410  0.0732  0.458
##
## $est_fs
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## D2D30 0.2996 0.0434      0 0.0463  0.2212  0.4048      0
## D2D40 0.3946 0.0754      0 0.0773  0.2554  0.5498      0
## D2D50 0.2663 0.0438      0 0.0485  0.1845  0.3858      0
##
## $p_iv
## [1] 3
##
## $N
## [1] 3350
##
## $N_cl
## [1] 282
##
## $df
## [1] 3316
##
## $nvalues
##      elecc_presid2013 ratio_treat D2D30 D2D40 D2D50
## [1,]                2           56    2    2    2
##
## attr("class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



Colantone and Stanig (2018)

Replication Summary

Unit of analysis	region*year
Treatment	regional import shock from China
Instrument	Chinese imports to the United States
Outcome	Economic nationalism
Model	Table1(1)

```
df <- readRDS("./rawdata/ajps_Colantone_etal_2018.rds")
D <- "import_shock"
Y <- "median_nationalism"
Z <- "instrument_for_shock"
controls <- NULL
cl <- "nuts2_year"
FE <- "fix_effect"
weights <- NULL
(g <- ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl, weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.6442 0.2934 2.1955 0.0691  1.2193 0.0281
## Boot.c   0.6442 0.3691 1.7452 0.1986  1.5873 0.0000
```

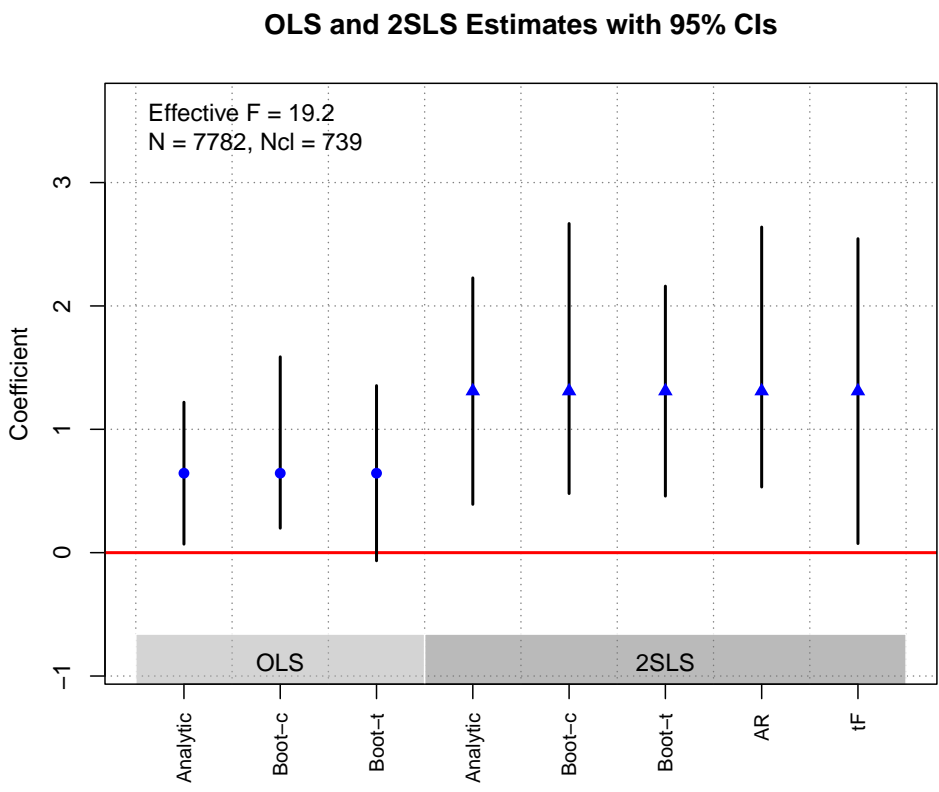
```

## Boot.t    0.6442 0.2934 2.1955 -0.0655    1.3539 0.0670
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 1.3096 0.4682 2.797 0.3919    2.2273 0.0052
## Boot.c   1.3096 0.5482 2.389 0.4795    2.6676 0.0020
## Boot.t   1.3096 0.4682 2.797 0.4588    2.1604 0.0090
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
## 10.9563    1.0000 7780.0000    0.0009
##
## $AR$ci.print
## [1] "[0.5323, 2.6393]"
##
## $AR$ci
## [1] 0.5323 2.6393
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 1810.3678    42.8350    19.1709    11.0506    19.1709
##
## $rho
## [1] 0.4358
##
## $tF
##          F      cF      Coef      SE      t CI2.5% CI97.5% p-value
## 19.1709 2.6386 1.3096 0.4682 2.7970 0.0741 2.5450 0.0377
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## instrument_for_shock 0.0514 0.0156 0.001 0.0192 0.0222 0.0945 0.002
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## instrument_for_shock 0.0392 0.009 0 0.0118 0.0258 0.0705 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 7782

```

```
##
## $N_cl
## [1] 739
##
## $df
## [1] 7724
##
## $nvalues
##      median_nationalism import_shock instrument_for_shock
## [1,]                167             739                   739
##
## attr("class")
## [1] "ivDiag"
```

plot_coef(g)



Coppock and Green (2016)

Replication Summary

Unit of analysis	individual
Treatment	voting in November 2007 municipal elections
Instrument	mailing showing 2005 Vote
Outcome	voting in the 2008 presidential primary
Model	Table2(2)

```
df<-readRDS("./rawdata/ajps_Coppock_etal_2016.rds")
D <-"og2007"
Y <- "JAN2008"
Z <- "treat2"
controls <- NULL
cl <- "hh"
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.3126 0.0014 229.6550 0.3099 0.3152      0
## Boot.c   0.3126 0.0014 223.0352 0.3097 0.3153      0
## Boot.t   0.3126 0.0014 229.6550 0.3106 0.3145      0
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.3728 0.0909 4.1013 0.1946 0.5509 0.000
## Boot.c   0.3728 0.0939 3.9688 0.1836 0.5504 0.002
## Boot.t   0.3728 0.0909 4.1013 0.2447 0.5008 0.000
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
## 15.4540 1.0000 773554.0000 0.0001
##
## $AR$ci.print
## [1] "[0.1946, 0.5564]"
##
## $AR$ci
## [1] 0.1946 0.5564
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 165.8659 151.8337 113.3680 102.8618 113.3680
##
## $rho
## [1] 0.0146
##
## $tF
##           F      cF      Coef      SE      t CI2.5% CI97.5% p-value
```

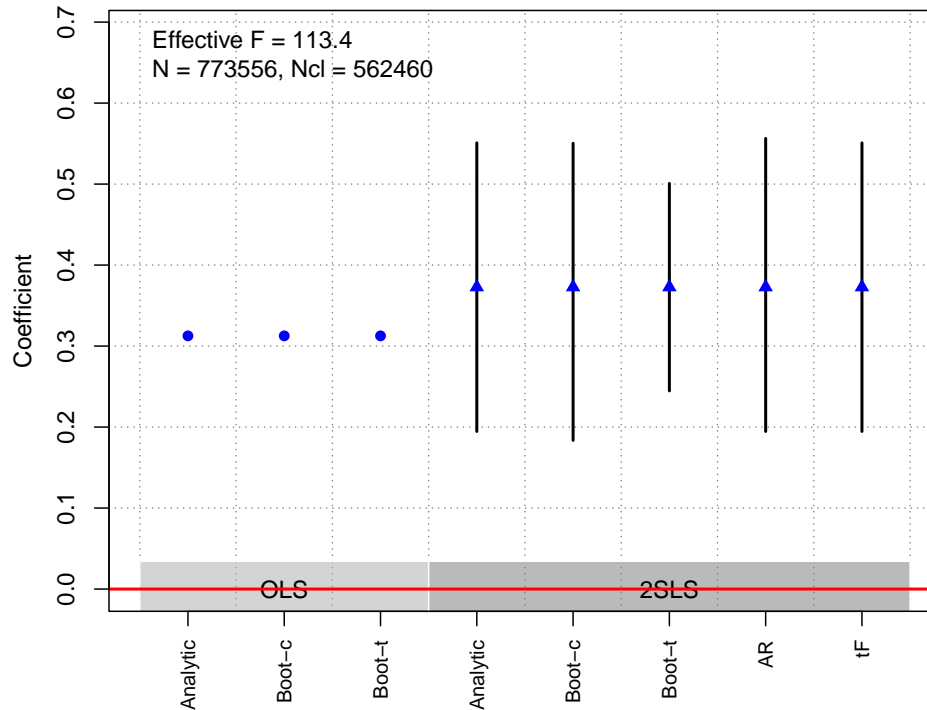
```

## 113.3680  1.9600  0.3728  0.0909  4.1013  0.1946  0.5509  0.0000
##
## $est_rf
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## treat2 0.0187 0.0048 1e-04 0.0049  0.009  0.0283  0.002
##
## $est_fs
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## treat2 0.0502 0.0047  0 0.0049  0.0398  0.0595  0
##
## $p_iv
## [1] 1
##
## $N
## [1] 773556
##
## $N_cl
## [1] 562460
##
## $df
## [1] 773554
##
## $nvalues
##      JAN2008 og2007 treat2
## [1,]      2      2      2
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



De La O (2013)

Replication Summary

Unit of analysis	village
Treatment	early coverage of Conditional Cash Transfer
Instrument	random assignment to early coverage
Outcome	incumbent party's vote share
Model	Table3(b1)

```
df <- readRDS("./rawdata/ajps_De_La_0_2013.rds")
D <- "early_progres_a_p"
Y <- "t2000"
Z <- "treatment"
controls <- c("avgpoverty", "pobtot1994", "votos_totales1994",
             "pri1994", "pan1994", "prd1994")
cl <- NULL
FE <- "villages"
weights <- NULL
(g <- ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
             cl =cl, weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0222 0.0466 0.4771 -0.0691  0.1136 0.6333
```

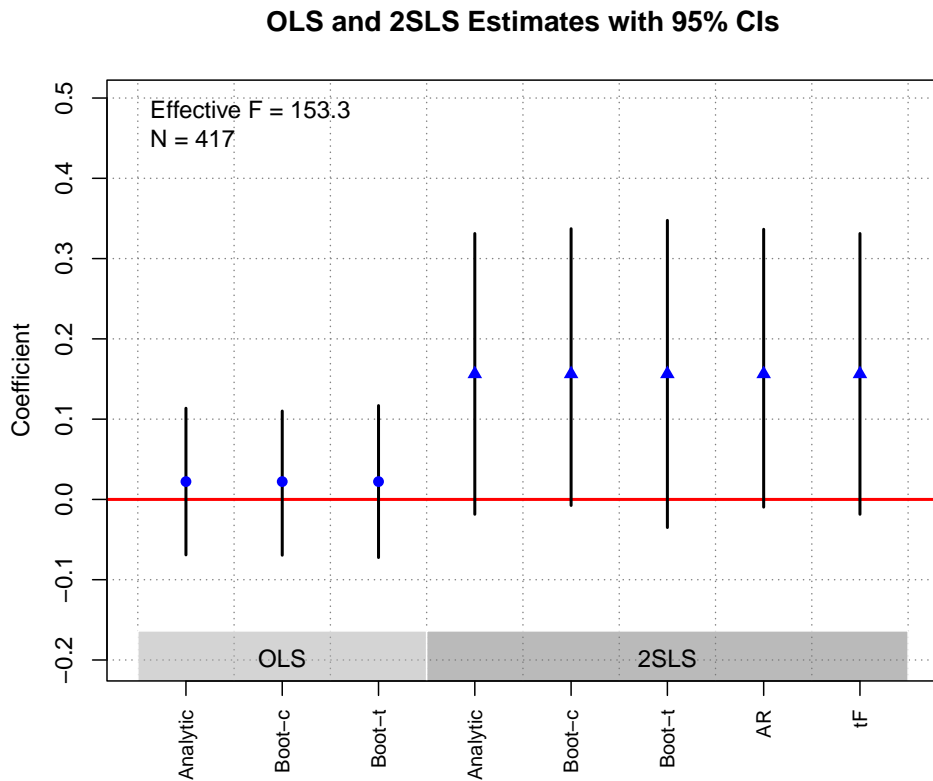
```

## Boot.c  0.0222 0.0460 0.4832 -0.0695  0.1101  0.7140
## Boot.t  0.0222 0.0466 0.4771 -0.0724  0.1169  0.6560
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.1563 0.0892 1.7521 -0.0185  0.3312  0.0798
## Boot.c   0.1563 0.0912 1.7132 -0.0075  0.3372  0.0740
## Boot.t   0.1563 0.0892 1.7521 -0.0350  0.3476  0.1020
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
##   3.3846   1.0000 415.0000  0.0665
##
## $AR$ci.print
## [1] "[-0.0096, 0.3365]"
##
## $AR$ci
## [1] -0.0096  0.3365
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
##   177.1916   153.2854         NA   144.9531   153.2854
##
## $rho
## [1] 0.556
##
## $tF
##          F      cF      Coef      SE      t  CI2.5% CI97.5% p-value
## 153.2854  1.9600  0.1563  0.0892  1.7521 -0.0185  0.3312  0.0798
##
## $est_rf
##          Coef      SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## treatment 0.0532 0.0296  0.0723 0.03  -0.0027  0.1096  0.074
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## treatment 0.3401 0.0275  0 0.0282  0.2871  0.3957  0
##
## $p_iv
## [1] 1
##
## $N

```

```
## [1] 417
##
## $N_cl
## NULL
##
## $df
## [1] 396
##
## $nvalues
##      t2000 early_progesa_p treatment
## [1,]  407             251         2
##
## attr("class")
## [1] "ivDiag"
```

plot_coef(g)



Goldstein and You (2017)

Replication Summary

Unit of analysis	city
Treatment	lobbying spending
Instrument	direct flight to Washington, DC
Outcome	total earmarks or grants awarded

Replication Summary

Model Table4(4)

```
df <- readRDS("./rawdata/ajps_Goldstein_etal_2017.rds")
df <- as.data.frame(df)
Y <- "ln_recovery"
D <- "ln_citylob"
Z <- c("direct_flight_dc", "diverge2_r")
controls <- c("pop_r", "land_r", "water_r", "senior_r", "student_r", "ethnic_r",
             "mincome_r", "unemp_r", "poverty_r", "gini_r", "city_propertytaxshare_r",
             "city_intgovrevenueshare_r", "city_airexp_r", "houdem_r", "ln_countylob")
cl <- "state2"
FE <- "state2"
weights <- NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl, weights=weights, cores = cores, parallel = TRUE))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0648 0.0208 3.1171 0.0240 0.1055 0.0018
## Boot.c   0.0648 0.0245 2.6485 0.0282 0.1231 0.0020
## Boot.t   0.0648 0.0208 3.1171 0.0245 0.1050 0.0010
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.476 0.1361 3.4987 0.2094 0.7427 0.0005
## Boot.c   0.476 0.1580 3.0121 0.1707 0.7760 0.0120
## Boot.t   0.476 0.1361 3.4987 0.2789 0.6732 0.0000
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    8.2957    2.0000 1259.0000 0.0003
##
## $AR$ci.print
## [1] "[0.1958, 0.9263]"
##
## $AR$ci
## [1] 0.1958 0.9263
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
```

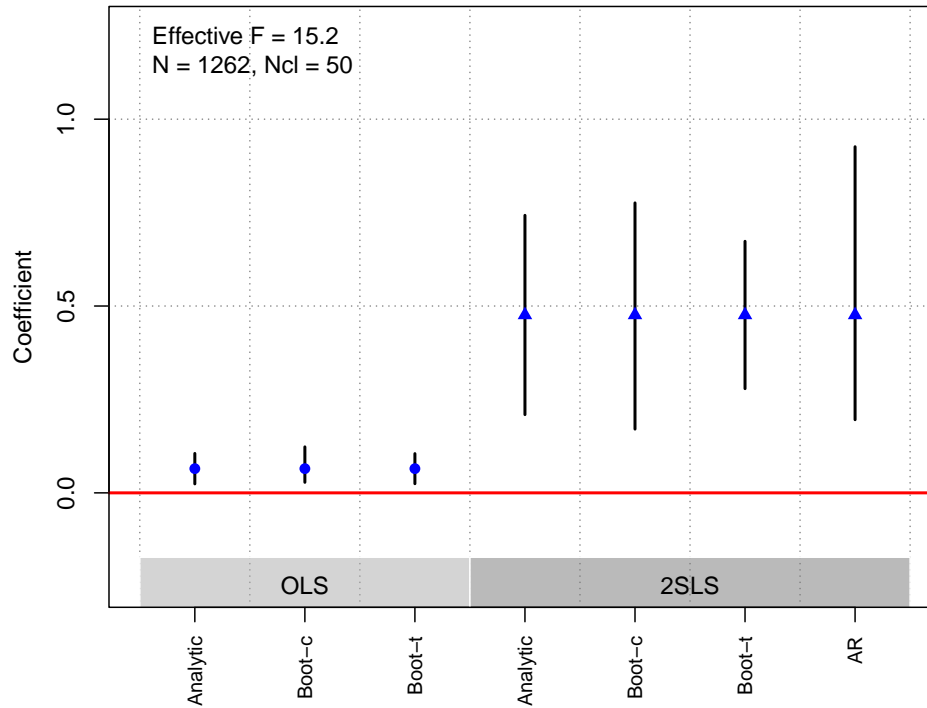
```

##      16.6195      13.7688      15.7426      14.5282      15.1587
##
## $rho
## [1] 0.1645
##
## $est_rf
##              Coef      SE p.value    SE.b CI.b2.5% CI.b97.5% p.value.b
## direct_flight_dc 1.2403 0.5428 0.0223 0.6128 -0.3190 2.1915 0.098
## diverge2_r      0.3010 0.1688 0.0745 0.1764 -0.0423 0.6681 0.070
##
## $est_fs
##              Coef      SE p.value    SE.b CI.b2.5% CI.b97.5% p.value.b
## direct_flight_dc 2.6658 0.7247 2e-04 0.7570 1.0869 4.0204 0.00
## diverge2_r      0.6070 0.2164 5e-03 0.2278 0.1954 1.0927 0.01
##
## $p_iv
## [1] 2
##
## $N
## [1] 1262
##
## $N_cl
## [1] 50
##
## $df
## [1] 49
##
## $nvalues
##      ln_recovery ln_citylob direct_flight_dc diverge2_r
## [1,]      1196      135                2      1262
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



Hager and Hilbig (2019) a

Replication Summary

Unit of analysis	city
Treatment	equitable inheritance customs
Instrument	mean elevation
Outcome	female representation
Model	Table3(1)

```
df<-readRDS("./rawdata/ajps_Hager_etal_2019.rds")
D <-"fair_dic"
Y <- "gem_women_share"
Z <- "elev_mean"
controls <- c("lon", "lat", "childlabor_mean_1898",
             "support_expenses_total_capita", "gem_council",
             "gem_pop_density", "pop_tot")
cl<- NULL
FE<- c("state2", "law_cat2")
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
```

```

## Analytic 0.0072 0.0042 1.7010 -0.0011 0.0155 0.0889
## Boot.c 0.0072 0.0042 1.7136 -0.0011 0.0148 0.0900
## Boot.t 0.0072 0.0042 1.7010 -0.0007 0.0151 0.0810
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.1363 0.0262 5.1939 0.0849 0.1878 0
## Boot.c 0.1363 0.0273 5.0006 0.0886 0.1933 0
## Boot.t 0.1363 0.0262 5.1939 0.0844 0.1883 0
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
## 38.9099 1.0000 3848.0000 0.0000
##
## $AR$ci.print
## [1] "[0.0901, 0.1957]"
##
## $AR$ci
## [1] 0.0901 0.1957
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 122.1930 79.2985 NA 76.7872 79.2985
##
## $rho
## [1] 0.1758
##
## $tF
##          F      cF      Coef      SE      t CI2.5% CI97.5% p-value
## 79.2985 2.0200 0.1363 0.0262 5.1939 0.0833 0.1894 0.0000
##
## $est_rf
##          Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## elev_mean -1e-04 0 0 0 -2e-04 -1e-04 0
##
## $est_fs
##          Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## elev_mean -9e-04 1e-04 0 1e-04 -0.0011 -7e-04 0
##
## $p_iv
## [1] 1
##

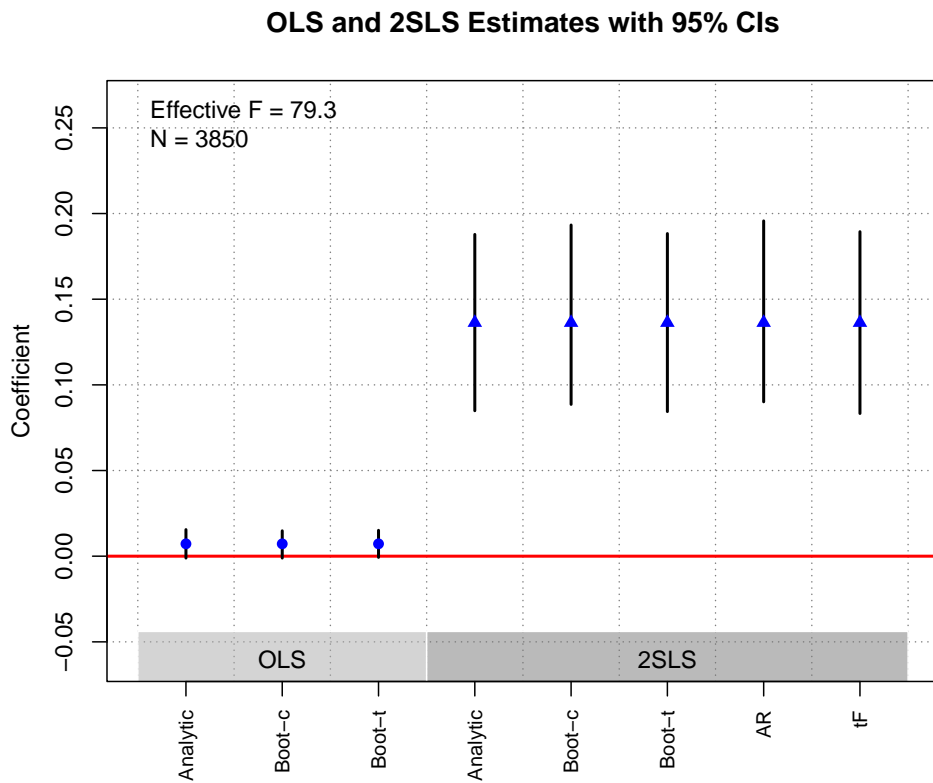
```

```

## $N
## [1] 3850
##
## $N_c1
## NULL
##
## $df
## [1] 3831
##
## $nvalues
##      gem_women_share fair_dic elev_mean
## [1,]           230         2       3850
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`



Hager and Hilbig (2019) b

Replication Summary

Unit of analysis	city
Treatment	equitable inheritance customs
Instrument	distance to rivers

Replication Summary

Outcome female representation
Model Table3(2)

```
df<-readRDS("./rawdata/ajps_Hager_etal_2019.rds")
D <-"fair_dic"
Y <- "gem_women_share"
Z <-"river_dist_min"
controls <- c("lon", "lat", "childlabor_mean_1898",
              "support_expenses_total_capita", "gem_council",
              "gem_pop_density", "pop_tot")
cl<- NULL
FE<- c("law_cat2")
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
            cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.015 0.0073 2.0379 0.0006 0.0293 0.0416
## Boot.c   0.015 0.0072 2.0791 0.0012 0.0288 0.0360
## Boot.t   0.015 0.0073 2.0379 0.0009 0.0290 0.0360
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0513 0.0239 2.1441 0.0044 0.0982 0.032
## Boot.c   0.0513 0.0244 2.1071 0.0070 0.1010 0.030
## Boot.t   0.0513 0.0239 2.1441 0.0048 0.0978 0.034
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
## 4.8070 1.0000 864.0000 0.0286
##
## $AR$ci.print
## [1] "[0.0058, 0.1006]"
##
## $AR$ci
## [1] 0.0058 0.1006
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
```

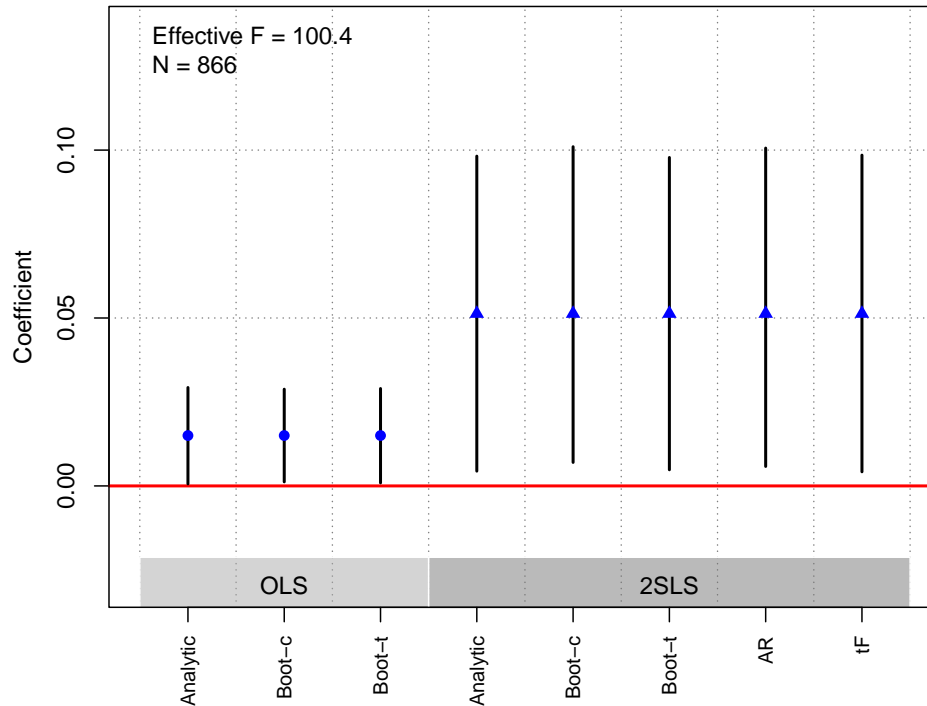
```

##      99.1676      100.3609          NA      95.5934      100.3609
##
## $rho
## [1] 0.3222
##
## $tF
##      F      cF      Coef      SE      t      CI2.5%      CI97.5%      p-value
## 100.3609  1.9700  0.0513  0.0239  2.1441  0.0042  0.0985  0.0329
##
## $est_rf
##              Coef      SE p.value      SE.b CI.b2.5% CI.b97.5% p.value.b
## river_dist_min -5e-04 2e-04  0.0291 2e-04  -0.001  -1e-04  0.03
##
## $est_fs
##              Coef      SE p.value      SE.b CI.b2.5% CI.b97.5% p.value.b
## river_dist_min -0.0105 0.001      0 0.0011  -0.0125  -0.0083  0
##
## $p_iv
## [1] 1
##
## $N
## [1] 866
##
## $N_c1
## NULL
##
## $df
## [1] 856
##
## $nvalues
##      gem_women_share fair_dic river_dist_min
## [1,]              110      2          866
##
## attr(,"class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



Hong et al. (2022)

Replication Summary

Unit of analysis	township
Treatment	NVM subsidy per voter
Instrument	Terrain elevation slope
Outcome	Park's vote share in 2012
Model	Table3(3)

```
df <-readRDS("./rawdata/ajps_Hong_etal_2022.rds")
df<-as.data.frame(df)
D<-"total_Lamount_1974_1978_perelect"
Y <- "E18ConsSh"
Z <- c("te_median1", "ts_median1")
controls <- c("area_1970", "demo_female_share_1966", "demo_age_15plus_1966",
              "demo_illiterate_1966", "demo_pop_ch_1970_1966", "E17ConsSh", "eup")
cl <- "CTY_cd"
FE <- "CTY_cd"
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
```

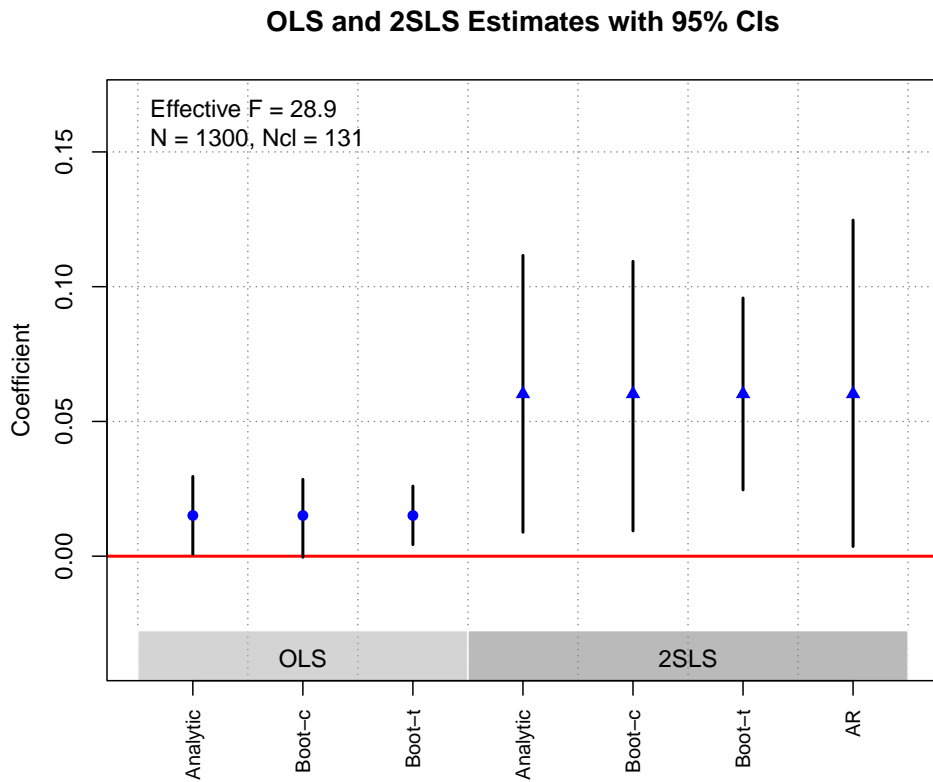
```

## Analytic 0.0151 0.0074 2.0600 0.0007 0.0296 0.0394
## Boot.c 0.0151 0.0073 2.0641 -0.0004 0.0285 0.0560
## Boot.t 0.0151 0.0074 2.0600 0.0043 0.0260 0.0090
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0602 0.0262 2.2980 0.0089 0.1116 0.0216
## Boot.c 0.0602 0.0257 2.3425 0.0094 0.1094 0.0220
## Boot.t 0.0602 0.0262 2.2980 0.0246 0.0958 0.0070
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
##    3.2888    2.0000 1297.0000 0.0376
##
## $AR$ci.print
## [1] "[0.0036, 0.1247]"
##
## $AR$ci
## [1] 0.0036 0.1247
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
##    34.7064    29.0832    28.2296    28.1134    28.8604
##
## $rho
## [1] 0.2376
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## te_median1 -0.0036 0.0233 0.8774 0.0235 -0.0518 0.0412 0.840
## ts_median1 0.0020 0.0010 0.0509 0.0010 0.0002 0.0039 0.028
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## te_median1 0.3276 0.1352 0.0154 0.1444 0.0443 0.6181 0.016
## ts_median1 0.0171 0.0061 0.0050 0.0062 0.0055 0.0296 0.002
##
## $p_iv
## [1] 2
##
## $N
## [1] 1300

```

```
##
## $N_cl
## [1] 131
##
## $df
## [1] 130
##
## $nvalues
##      E18ConsSh total_Lamount_1974_1978_perelect te_median1 ts_median1
## [1,]      1292                1285          1300          1232
##
## attr("class")
## [1] "ivDiag"
```

`plot_coef(g)`



Kim (2019)

Replication Summary

Unit of analysis	municipality*year
Treatment	Democratic institutions
Instrument	population threshold
Outcome	women political engagement
Model	Table2(1)

```
df<- readRDS("../rawdata/ajps_Kim_2019.rds")
D <-"direct"
Y <- "wm_turnout"
Z <- "new"
controls <- c("left", "wm_voters", "enep")
cl <- NULL
FE <- "year"
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.017 0.4897 0.0346 -0.9429  0.9768  0.9724
## Boot.c   0.017 0.5039 0.0337 -0.9780  1.0297  0.8540
## Boot.t   0.017 0.4897 0.0346 -0.9866  1.0205  0.9630
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 3.9287 1.0855 3.6192  1.8011  6.0563  3e-04
## Boot.c   3.9287 1.1465 3.4267  1.9919  6.5383  0e+00
## Boot.t   3.9287 1.0855 3.6192  1.6799  6.1774  0e+00
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
## 14.3152  1.0000 2747.0000  0.0002
##
## $AR$ci.print
## [1] "[1.8662, 6.0997]"
##
## $AR$ci
## [1] 1.8662 6.0997
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
## 1007.3382  914.6461      NA      955.6475  914.6461
##
## $rho
## [1] 0.5186
##
## $tF
##          F      cF      Coef      SE      t  CI2.5%  CI97.5%  p-value
```

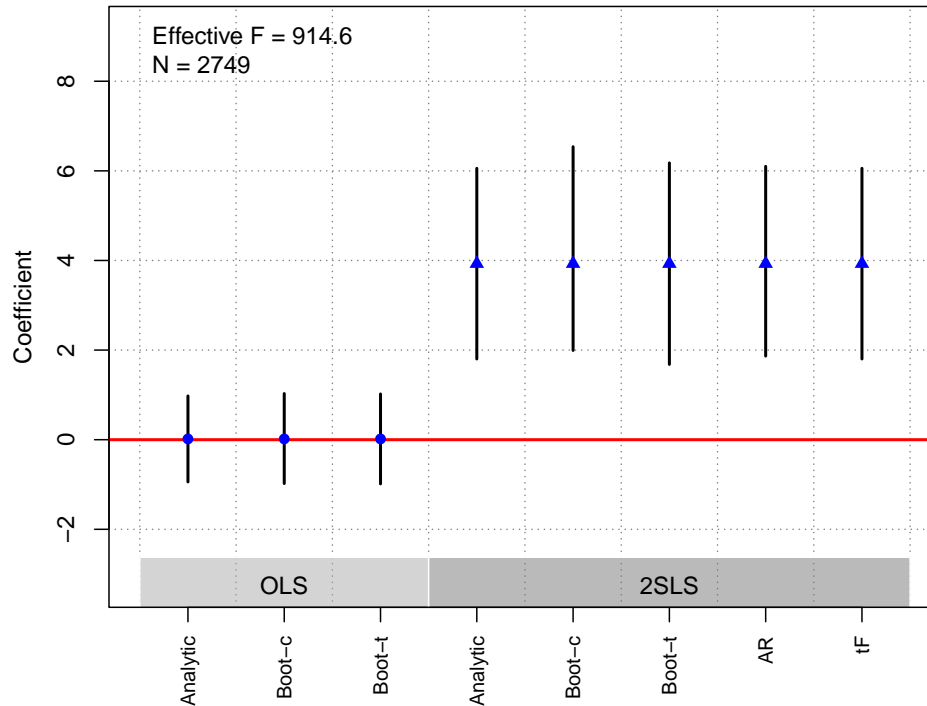
```

## 914.6461  1.9600  3.9287  1.0855  3.6192  1.8011  6.0563  0.0003
##
## $est_rf
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## new 1.949 0.516  2e-04 0.5387  0.9879  3.0959  0
##
## $est_fs
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## new 0.4961 0.0164  0 0.016  0.4585  0.522  0
##
## $p_iv
## [1] 1
##
## $N
## [1] 2749
##
## $N_c1
## NULL
##
## $df
## [1] 2738
##
## $nvalues
##      wm_turnout direct new
## [1,]      2606      2  2
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



Kocher et al. (2011)

Replication Summary

Unit of analysis	hamlet (smallest population unit)
Treatment	aerial bombing
Instrument	past insurgent control
Outcome	changes in local control
Model	Table5(5B)

```
df<-readRDS("./rawdata/ajps_Kocher_etal_2011.rds")
D <- "bombed_969"
Y<- "mod2a_1adec"
Z <- c("mod2a_1ajul", "mod2a_1aaug")
controls <- c("mod2a_1asep", "score", "ln_dist", "std", "lnhpop")
cl<- NULL
FE <-NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0249 0.0042 5.8926 0.0166 0.0332 0
## Boot.c   0.0249 0.0043 5.7687 0.0178 0.0349 0
```

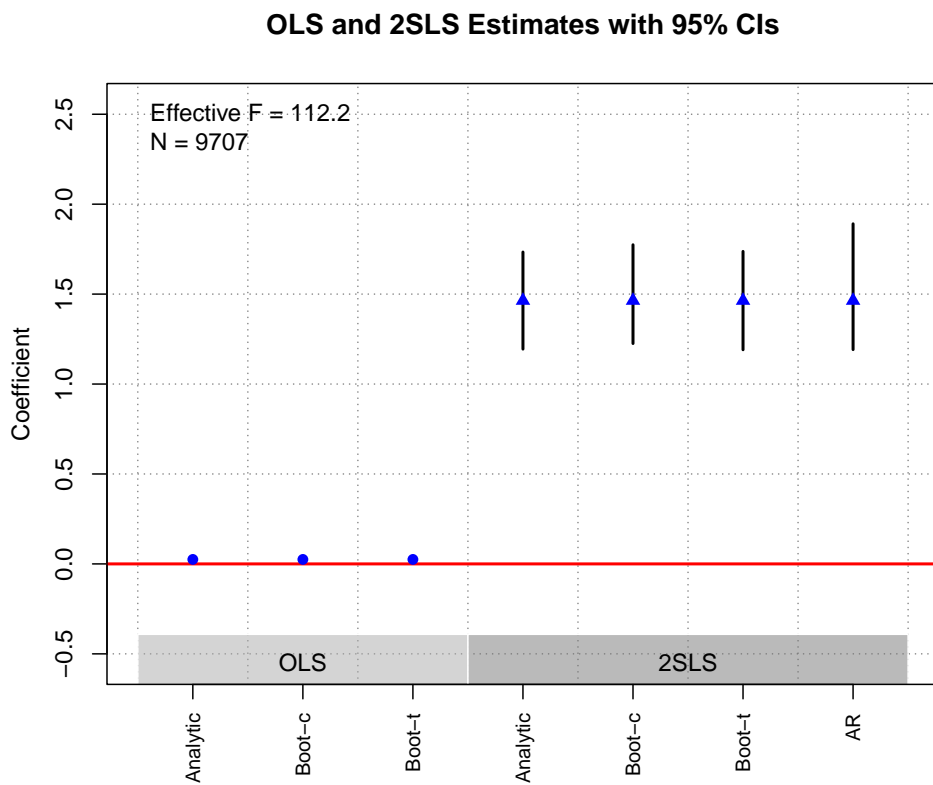
```

## Boot.t    0.0249 0.0042 5.8926  0.0163  0.0335      0
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 1.464 0.1377 10.6345  1.1942  1.7339      0
## Boot.c   1.464 0.1382 10.5902  1.2253  1.7743      0
## Boot.t   1.464 0.1377 10.6345  1.1906  1.7374      0
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
## 681.5407  2.0000 9704.0000  0.0000
##
## $AR$ci.print
## [1] "[1.1914, 1.8908]"
##
## $AR$ci
## [1] 1.1914 1.8908
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
## 44.1703    59.8861      NA      60.8767    112.1923
##
## $rho
## [1] 0.095
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## mod2a_1ajul 0.2562 0.0123      0 0.0118  0.2330  0.2779      0
## mod2a_1aug  0.1830 0.0134      0 0.0131  0.1577  0.2089      0
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## mod2a_1ajul 0.1681 0.0284      0 0.0283  0.1126  0.2243      0
## mod2a_1aug  0.1328 0.0311      0 0.0313  0.0752  0.1965      0
##
## $p_iv
## [1] 2
##
## $N
## [1] 9707
##
## $N_c1

```

```
## NULL
##
## $df
## [1] 9700
##
## $nvalues
##      mod2a_1adec bombed_969 mod2a_1ajul mod2a_1aaug
## [1,]          5          35          5          5
##
## attr(,"class")
## [1] "ivDiag"
```

plot_coef(g)



Lelkes et al. (2017)

Replication Summary

Unit of analysis	state*year
Treatment	number of broadband Internet providers
Instrument	state-level ROW index
Outcome	affective polarization
Model	Table1(3)

```

df<-readRDS("./rawdata/ajps_Lelkes_2017.rds")
D <-"D"
Y <- "outcome"
Z <- "Total_log"
controls <- c("region", "percent_black", "percent_white",
              "percent_male", "lowed", "unemploymentrate",
              "density", "HHINC_log")
cl<- "state"
FE <- "year"
weights=NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))

```

```

## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0041 0.0031 1.3481 -0.0019  0.0102  0.1776
## Boot.c   0.0041 0.0037 1.1060 -0.0024  0.0121  0.2160
## Boot.t   0.0041 0.0031 1.3481 -0.0011  0.0094  0.1250
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0316 0.0141 2.2364  0.0039  0.0593  0.0253
## Boot.c   0.0316 0.1224 0.2580 -0.0055  0.1415  0.0700
## Boot.t   0.0316 0.0141 2.2364  0.0097  0.0534  0.0060
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##      4.6542      1.0000 114801.0000      0.0310
##
## $AR$ci.print
## [1] "[0.0036, 0.0731]"
##
## $AR$ci
## [1] 0.0036 0.0731
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
## 9525.8467  8161.7346    11.1632    7.4361    11.1632
##
## $rho
## [1] 0.2768
##

```

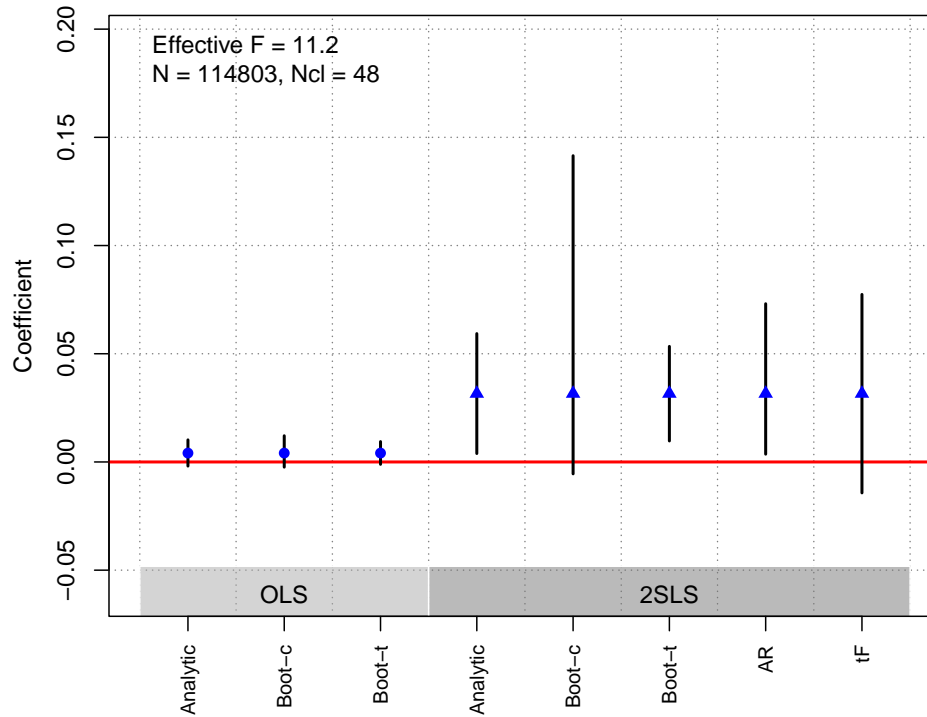
```

## $tF
##      F      cF      Coef      SE      t  CI2.5% CI97.5% p-value
## 11.1632 3.2489 0.0316 0.0141 2.2364 -0.0143 0.0774 0.1773
##
## $est_rf
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## Total_log 0.0033 0.0015 0.031 0.002 -1e-04 0.0077 0.056
##
## $est_fs
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## Total_log 0.1042 0.0312 8e-04 0.0382 0.012 0.1641 0.014
##
## $p_iv
## [1] 1
##
## $N
## [1] 114803
##
## $N_cl
## [1] 48
##
## $df
## [1] 114790
##
## $nvalues
##      outcome  D Total_log
## [1,] 2423 1438 43
##
## attr("class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



López-Moctezuma et al. (2020)

Replication Summary

Unit of analysis	individual
Treatment	town-hall meetings
Instrument	assignment to treatment
Outcome	voting behavior
Model	figure3(2)

```
df <-readRDS("./rawdata/ajps_Moctezuma_etal_2020.rds")
df<-as.data.frame(df)
D<-"treatment"
Y <- "vote"
Z <- "assignment"
  controls <- NULL
cl <- "barangay"
FE <- "city"
weights<-"weight.att"
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 16.1643 2.5956 6.2275 11.0769 21.2517 0.0000
```

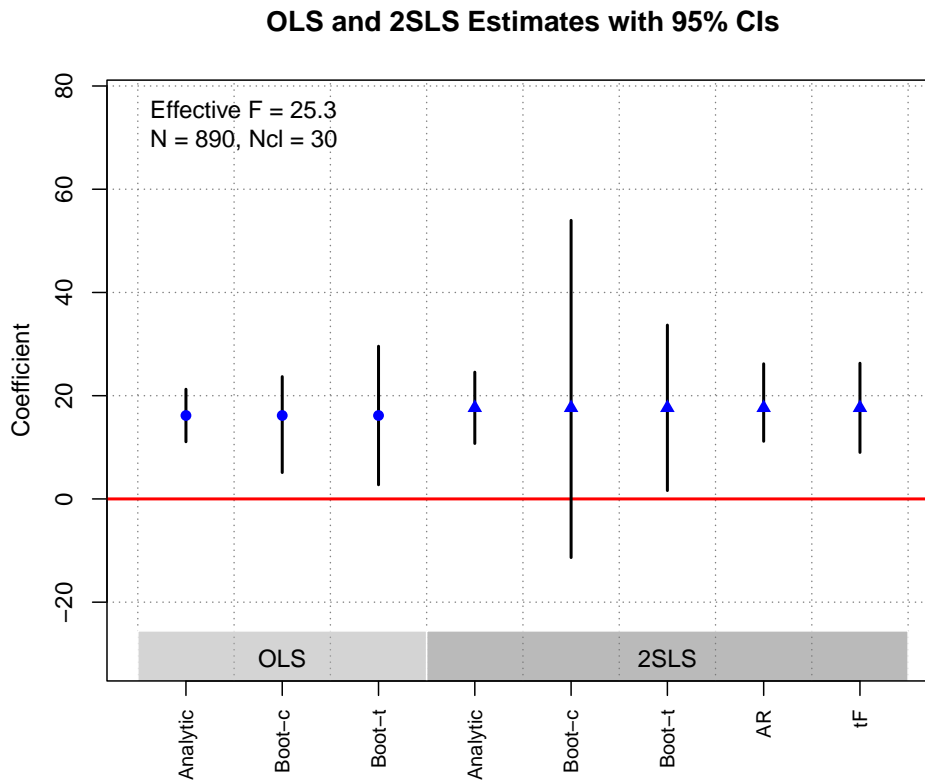
```

## Boot.c  16.1643 4.8128 3.3586  5.1162 23.6906  0.0040
## Boot.t  16.1643 2.5956 6.2275  2.7462 29.5823  0.0391
##
## $est_2sls
##           Coef      SE      t  CI 2.5% CI 97.5% p.value
## Analytic 17.6531  3.5231 5.0106 10.7478 24.5584  0.0000
## Boot.c   17.6531 85.4309 0.2066 -11.3544 53.9644  0.0701
## Boot.t   17.6531  3.5231 5.0106  1.6254 33.6807  0.0451
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
## 18.6344  1.0000 888.0000  0.0000
##
## $AR$ci.print
## [1] "[11.1705, 26.1790]"
##
## $AR$ci
## [1] 11.1705 26.1790
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
## 1663.9064  521.4034  25.2694  5.6018  25.2694
##
## $rho
## [1] 0.8089
##
## $tF
##           F      cF      Coef      SE      t  CI2.5% CI97.5% p-value
## 25.2694  2.4519 17.6531  3.5231  5.0106  9.0146 26.2915  0.0001
##
## $est_rf
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## assignment 13.2179 3.0776  0 5.9251  0.8983  24.6291  0.0321
##
## $est_fs
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## assignment 0.7488 0.149  0 0.3164 -0.021  1  0.0621
##
## $p_iv
## [1] 1
##
## $N

```

```
## [1] 890
##
## $N_cl
## [1] 30
##
## $df
## [1] 879
##
## $nvalues
##      vote treatment assignment
## [1,]  2          2          2
##
## attr("class")
## [1] "ivDiag"
```

plot_coef(g)



McClendon (2014)

Replication Summary

Unit of analysis	individual
Treatment	reading social esteem promising email
Instrument	assignment to treatment
Outcome	participation in LGBTQ events

Replication Summary

Model Table2(1)

```
df <- readRDS("./rawdata/ajps_McClendon_2014.rds")
D<-"openedesteem"
Y<- "intended"
Z <- "esteem"
controls <- NULL
cl<- NULL
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.2823 0.0339 8.3291 0.2159 0.3488 0
## Boot.c   0.2823 0.0347 8.1447 0.2163 0.3513 0
## Boot.t   0.2823 0.0339 8.3291 0.2135 0.3511 0
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.3149 0.0890 3.5376 0.1404 0.4893 4e-04
## Boot.c   0.3149 0.0896 3.5126 0.1470 0.4926 0e+00
## Boot.t   0.3149 0.0890 3.5376 0.1414 0.4883 4e-03
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
## 11.9462 1.0000 3645.0000 0.0006
##
## $AR$ci.print
## [1] "[0.1404, 0.4911]"
##
## $AR$ci
## [1] 0.1404 0.4911
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 103.7604 207.1798 NA 204.8498 207.1798
##
## $rho
```

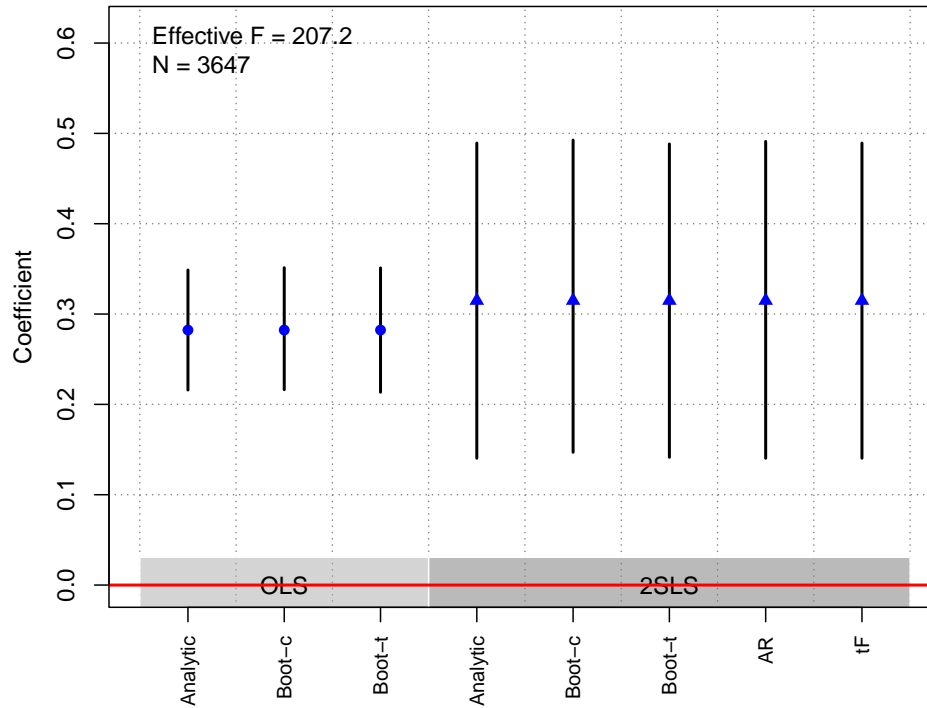
```

## [1] 0.1664
##
## $tF
##      F      cF      Coef      SE      t      CI2.5%  CI97.5%  p-value
## 207.1798  1.9600  0.3149  0.0890  3.5376  0.1404  0.4893  0.0004
##
## $est_rf
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## esteem 0.0247 0.0072 5e-04 0.0071 0.0105 0.0388 0
##
## $est_fs
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## esteem 0.0786 0.0055 0 0.0055 0.0683 0.0894 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 3647
##
## $N_c1
## NULL
##
## $df
## [1] 3645
##
## $nvalues
##      intended openedesteem esteem
## [1,]      2      2      2
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



Rueda (2017)

Replication Summary

Unit of analysis	city
Treatment	actual polling place size.
Instrument	the size of the polling station
Outcome	citizens' reports of electoral manipulation
Model	Table5(1)

```
df <- readRDS("./rawdata/ajps_Rueda_2017.rds")
D <- "lm_pob_mesa"
Y <- "e_vote_buying"
Z <- "lz_pob_mesa_f"
controls <- c("lpopulation", "lpotencial")
cl <- "muni_code"
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.675 0.1011 -6.6803 -0.8731 -0.4770      0
## Boot.c   -0.675 0.1026 -6.5778 -0.8883 -0.4888      0
```

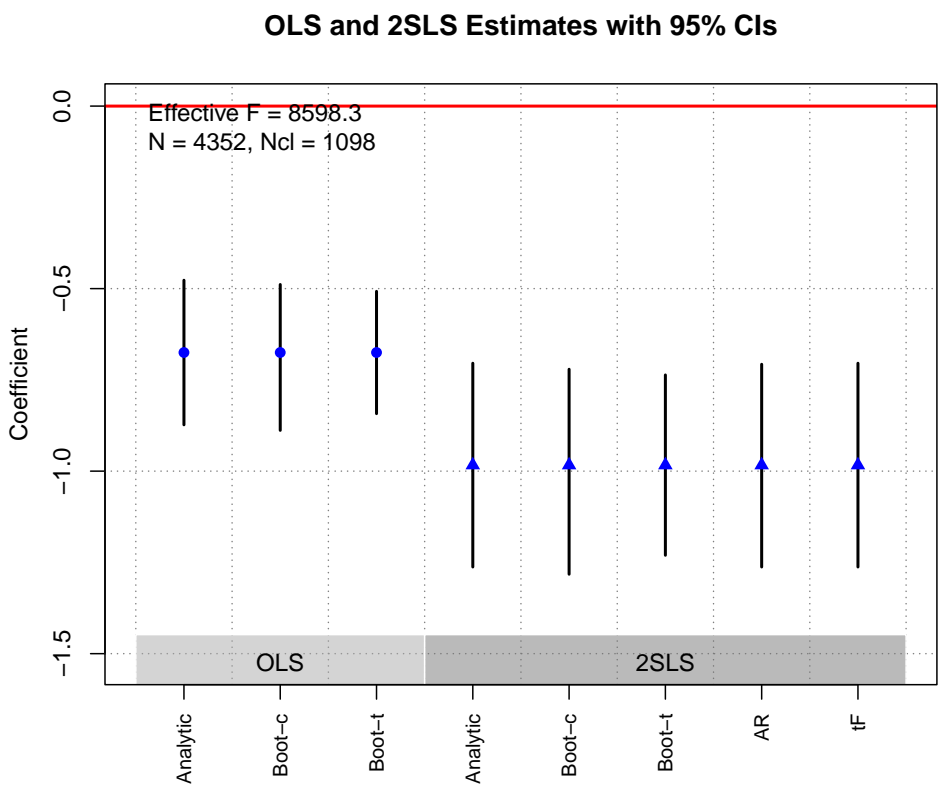
```

## Boot.t   -0.675 0.1011 -6.6803 -0.8425 -0.5075      0
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.9835 0.1424 -6.9071 -1.2626 -0.7044      0
## Boot.c   -0.9835 0.1481 -6.6425 -1.2824 -0.7209      0
## Boot.t   -0.9835 0.1424 -6.9071 -1.2305 -0.7365      0
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
## 48.4768  1.0000 4350.0000  0.0000
##
## $AR$ci.print
## [1] "[-1.2626, -0.7073]"
##
## $AR$ci
## [1] -1.2626 -0.7073
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
## 3106.387  3108.591  8598.326  8520.475  8598.326
##
## $rho
## [1] 0.6455
##
## $tF
##           F      cF      Coef      SE      t      CI2.5%  CI97.5%  p-value
## 8598.3264  1.9600  -0.9835  0.1424  -6.9071  -1.2626  -0.7044  0.0000
##
## $est_rf
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## lz_pob_mesa_f -0.7826 0.1124      0 0.1168  -1.024  -0.5756      0
##
## $est_fs
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## lz_pob_mesa_f 0.7957 0.0086      0 0.0086  0.7779  0.8107      0
##
## $p_iv
## [1] 1
##
## $N
## [1] 4352

```

```
##
## $N_cl
## [1] 1098
##
## $df
## [1] 4348
##
## $nvalues
##      e_vote_buying lm_pob_mesa lz_pob_mesa_f
## [1,]           16          4118          3860
##
## attr("class")
## [1] "ivDiag"
```

plot_coef(g)



Sexton et al. (2019)

Replication Summary

Unit of analysis	department*year
Treatment	health budget
Instrument	soldier fatalities
Outcome	health social service
Model	Table3(1)

```

df <-readRDS("./rawdata/ajps_Sexton_etal_2019.rds")
D<-"socialservice_b"
Y <- "Finfant_mortality"
Z <- "Lgk_budget"
controls <- c("Lgk_prebudget", "ln_pbi_pc", "execution_nohealth")
cl <- "deptcode"
FE <- c("year","deptcode")
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))

```

```

## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -1.3472 1.0152 -1.3270 -3.3371  0.6426  0.1845
## Boot.c   -1.3472 1.1374 -1.1845 -3.3914  1.2813  0.2589
## Boot.t   -1.3472 1.0152 -1.3270 -3.0501  0.3556  0.0948
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -15.0645  8.0376 -1.8743 -30.8181  0.6892  0.0609
## Boot.c   -15.0645 28.0180 -0.5377 -53.3001 11.7851  0.2263
## Boot.t   -15.0645  8.0376 -1.8743 -70.6769 40.5480  0.1855
##
## $AR
## $AR$Fstat
##      F      df1      df2      p
## 18.0386 1.0000 70.0000 0.0001
##
## $AR$ci.print
## [1] "[-66.3101, -5.4194]"
##
## $AR$ci
## [1] -66.3101 -5.4194
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
##      1.0172      2.5692      7.4923      2.7847      7.4923
##
## $rho
## [1] 0.1538
##
## $tF
##      F      cF      Coef      SE      t  CI2.5%  CI97.5%  p-value

```

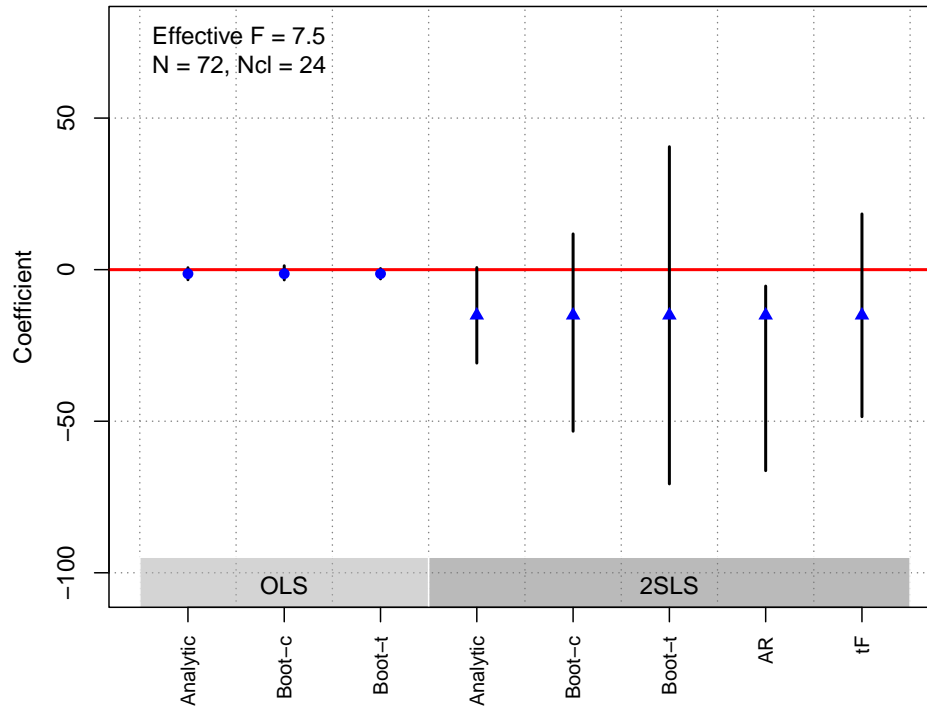
```

## 7.4923 4.1607 -15.0645 8.0376 -1.8743 -48.5065 18.3775 0.3773
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## Lgk_budget 4.3552 1.0481      0 2.1668 -1.5721 6.0768 0.1998
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## Lgk_budget -0.2891 0.1056 0.0062 0.1732 -0.698 -0.0156 0.0469
##
## $p_iv
## [1] 1
##
## $N
## [1] 72
##
## $N_c1
## [1] 24
##
## $df
## [1] 23
##
## $nvalues
##      Infant_mortality socialservice_b Lgk_budget
## [1,]                39                72                6
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



Spenkuch and Tillmann (2018)

Replication Summary

Unit of analysis	electoral district
Treatment	religion of voters living in the same areas more than three and a half centuries later
Instrument	individual princes' decisions concerning whether to adopt Protestantism
Outcome	Nazi vote share
Model	Table2(B1)

```
df <- readRDS("./rawdata/ajps_Spenkuch_etal_2018.rds")
D <- "r_1925C_kath"
Y <- "r_NSDAP_NOV1932_p"
Z <- c("r_kath1624", "r_gem1624")
controls <- c("r_1925C_juden", "r_1925C_others",
             "r_M1925C_juden", "r_M1925C_others")
cl <- 'WKNR'
FE <- NULL
weights="r_wahlberechtigte_NOV1932"
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
```

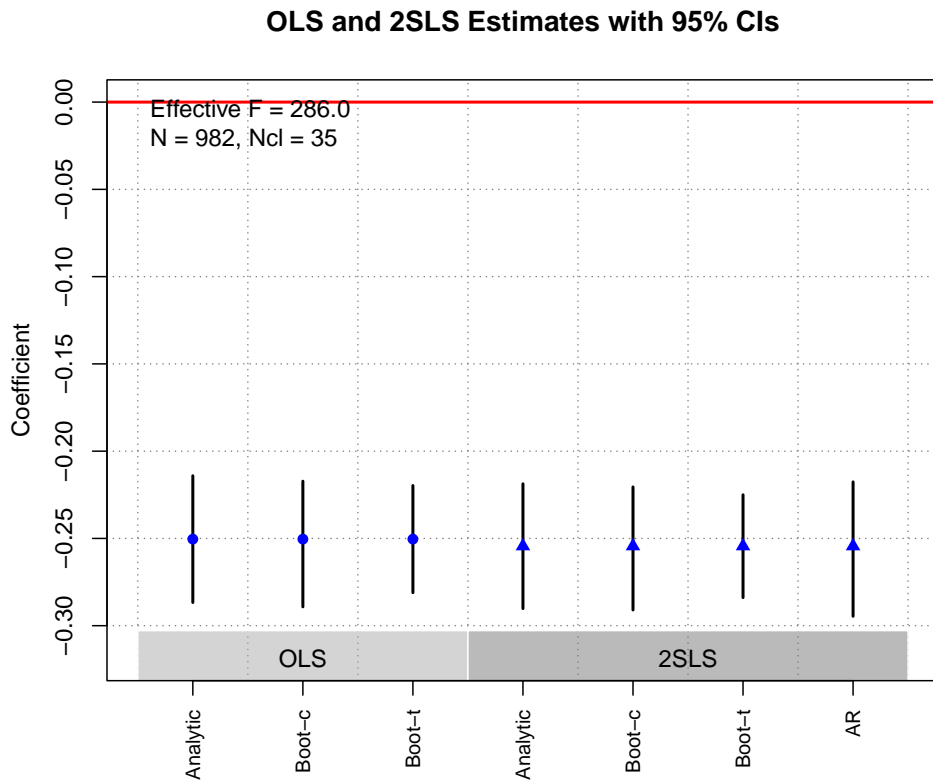
```

## Analytic -0.2504 0.0185 -13.5112 -0.2867 -0.2141 0
## Boot.c -0.2504 0.0188 -13.2921 -0.2892 -0.2172 0
## Boot.t -0.2504 0.0185 -13.5112 -0.2811 -0.2197 0
##
## $est_2sls
##          Coef      SE          t CI 2.5% CI 97.5% p.value
## Analytic -0.2544 0.0182 -13.9439 -0.2902 -0.2187 0
## Boot.c   -0.2544 0.0181 -14.0506 -0.2910 -0.2205 0
## Boot.t   -0.2544 0.0182 -13.9439 -0.2839 -0.2250 0
##
## $AR
## $AR$Fstat
##          F          df1          df2          p
## 89.3425  2.0000 979.0000  0.0000
##
## $AR$ci.print
## [1] "[-0.2946, -0.2176]"
##
## $AR$ci
## [1] -0.2946 -0.2176
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
## 1215.3547  726.7058  212.7390  221.4961  286.0263
##
## $rho
## [1] 0.8446
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## r_kath1624 -17.2028 1.2929 0 1.3039 -19.7471 -14.7375 0
## r_gem1624  -9.1477 1.5382 0 1.6096 -12.9046 -6.4089 0
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## r_kath1624 66.6657 3.232 0 3.1676 59.8434 72.5125 0
## r_gem1624 39.2697 4.320 0 4.5811 31.0940 50.0754 0
##
## $p_iv
## [1] 2
##
## $N
## [1] 982

```

```
##
## $N_cl
## [1] 35
##
## $df
## [1] 978
##
## $nvalues
##      r_NSDAP_NOV1932_p r_1925C_kath r_kath1624 r_gem1624
## [1,]                982          977          2          2
##
## attr("class")
## [1] "ivDiag"
```

plot_coef(g)



Stokes (2016)

Replication Summary

Unit of analysis	precinct
Treatment	turbine location
Instrument	wind speed
Outcome	vote turnout
Model	Table2(2)

```

df<-readRDS("./rawdata/ajps_Stokes_2016.rds")
D <-"prop_3km"
Y <- "chng_lib"
Z <- "avg_pwr_log"
controls <- c("mindistlake", "mindistlake_sq", "longitude",
              "long_sq", "latitude", "lat_sq", "long_lat")
cl <- NULL
FE <- "ed_id"
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
            cl =cl,weights=weights, cores = cores))

```

```

## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.0203 0.0073 -2.7638 -0.0347 -0.0059 0.0057
## Boot.c   -0.0203 0.0073 -2.7678 -0.0332 -0.0061 0.0100
## Boot.t   -0.0203 0.0073 -2.7638 -0.0345 -0.0061 0.0070
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.077 0.0282 -2.7289 -0.1323 -0.0217 0.0064
## Boot.c   -0.077 0.0305 -2.5217 -0.1384 -0.0194 0.0080
## Boot.t   -0.077 0.0282 -2.7289 -0.1324 -0.0216 0.0050
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    7.6582    1.0000  706.0000  0.0058
##
## $AR$ci.print
## [1] "[-0.1345, -0.0234]"
##
## $AR$ci
## [1] -0.1345 -0.0234
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
##    67.9032    65.7306         NA     61.7780     65.7306
##
## $rho
## [1] 0.3025
##
## $tF

```

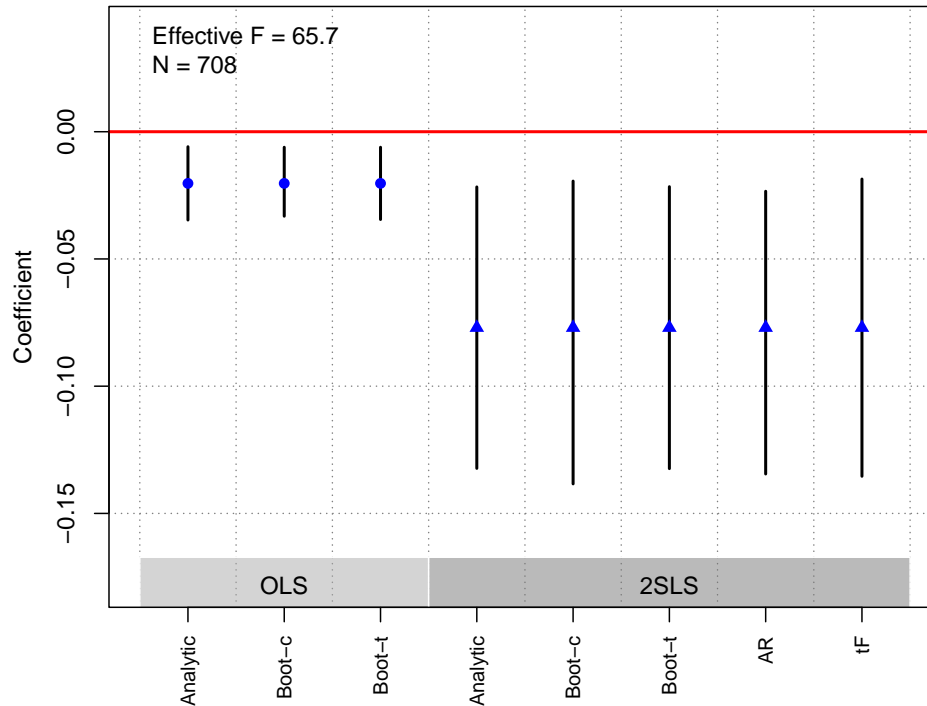
```

##      F      cF      Coef      SE      t  CI2.5% CI97.5% p-value
## 65.7306 2.0693 -0.0770 0.0282 -2.7289 -0.1354 -0.0186 0.0097
##
## $est_rf
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## avg_pwr_log -0.0585 0.0216 0.0069 0.0217 -0.1006 -0.0146 0.008
##
## $est_fs
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## avg_pwr_log 0.7602 0.0938 0 0.0967 0.5532 0.9273 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 708
##
## $N_cl
## NULL
##
## $df
## [1] 674
##
## $nvalues
##      chng_lib prop_3km avg_pwr_log
## [1,]      708      2      708
##
## attr("class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



Tajima (2013)

Replication Summary

Unit of analysis	village and urban neighborhood
Treatment	distance to police posts (as a proxy for exposure to military intervention)
Instrument	distance to health station
Outcome	incidence of communal violence
Model	Table1(4)

```
df<-readRDS("./rawdata/ajps_Tajima_2013.rds")
D <- "z2_distpospol"
Y <- "horiz2"
Z <- "z2_dispuskes"
controls <- c("flat", "z2_altitude", "urban", "natres", "z2_logvillpop", "z2_logdensvil",
             "z2_povrateksvil", "z2_fgtskvid", "z2_covyredvil", "z2_npwperhh",
             "z2_ethfractvil", "z2_ethfractsd", "z2_ethfractd", "z2_relfractvil",
             "z2_relfractsd", "z2_relfractd", "z2_ethclustsd", "z2_ethclustvd",
             "z2_relclustsd", "z2_relclustvd", "z2_wgcovegvil", "z2_wgcovegsd",
             "z2_wgcovegd", "z2_wgcovrgvil", "z2_wgcovrgsd", "z2_wgcovrgd",
             "natdis", "japanese_off_java", "islam", "split_kab03", "split_vil03")
cl <- 'kavid03'
FE <- 'prop'
weights<-"probit_touse_wts03"
```

```
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
cl =cl,weights=weights, cores = cores))
```

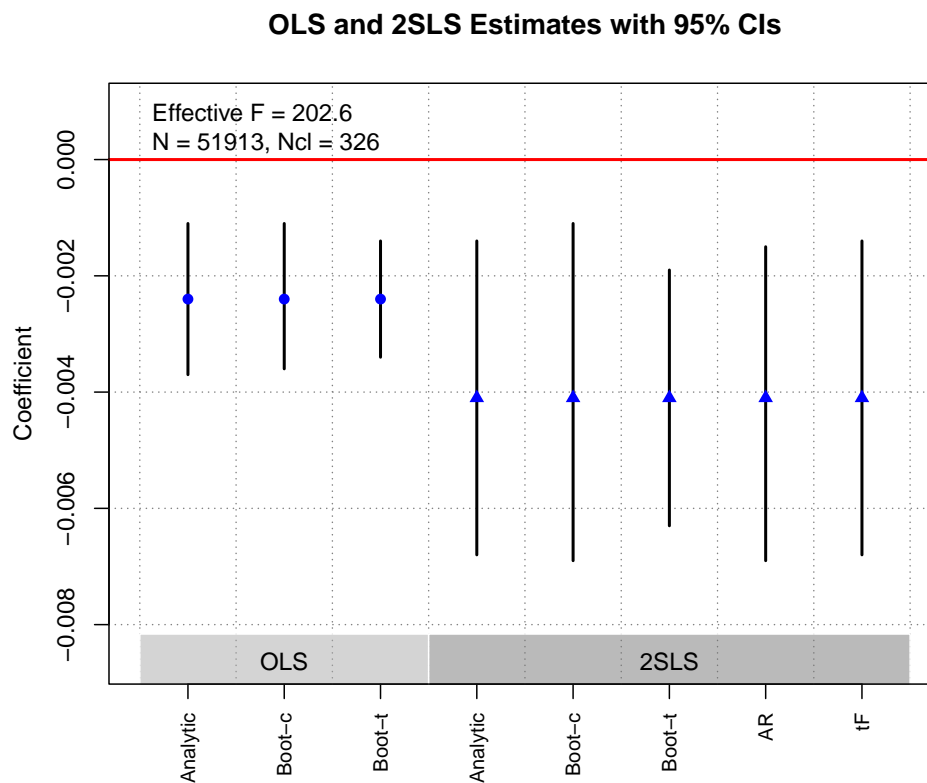
```
## $est_ols
##           Coef      SE          t CI 2.5% CI 97.5% p.value
## Analytic -0.0024 6e-04 -3.7223 -0.0037 -0.0011 2e-04
## Boot.c   -0.0024 7e-04 -3.6349 -0.0036 -0.0011 0e+00
## Boot.t   -0.0024 6e-04 -3.7223 -0.0034 -0.0014 0e+00
##
## $est_2sls
##           Coef      SE          t CI 2.5% CI 97.5% p.value
## Analytic -0.0041 0.0014 -3.0103 -0.0068 -0.0014 0.0026
## Boot.c   -0.0041 0.0015 -2.7409 -0.0069 -0.0011 0.0140
## Boot.t   -0.0041 0.0014 -3.0103 -0.0063 -0.0019 0.0000
##
## $AR
## $AR$Fstat
##           F          df1          df2          p
##      9.0632      1.0000 51911.0000      0.0026
##
## $AR$ci.print
## [1] "[-0.0069, -0.0015]"
##
## $AR$ci
## [1] -0.0069 -0.0015
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
## 13363.7649  1529.0807   202.6374   197.8245    202.6374
##
## $rho
## [1] 0.4527
##
## $tF
##           F          cF          Coef          SE          t  CI2.5%  CI97.5%  p-value
## 202.6374  1.9600  -0.0041  0.0014  -3.0103  -0.0068  -0.0014  0.0026
##
## $est_rf
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## z2_dispuskes -0.0019 6e-04 0.0026 7e-04 -0.003 -5e-04 0.014
##
## $est_fs
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
```

```

## z2_dispuskes 0.447 0.0314      0 0.0318  0.3832  0.5074      0
##
## $p_iv
## [1] 1
##
## $N
## [1] 51913
##
## $N_cl
## [1] 326
##
## $df
## [1] 51853
##
## $nvalues
##      horiz2 z2_distpospol z2_dispuskes
## [1,]      2          101          101
##
## attr("class")
## [1] "ivDiag"

```

```
plot_coef(g)
```



Trounstine (2016)

Replication Summary

Unit of analysis	city*year
Treatment	racial segregation
Instrument	the number of waterways in a city; logged population
Outcome	direct general expenditures
Model	Table5(1)

```
df<-readRDS("./rawdata/ajps_Trounstine_2016.rds")
D <-"H_citytract_NHW_i"
Y <- "dgepercap_cpi"
Z <- c("total_rivs_all", "logpop")
controls <- c("dgepercap_cpilag", "diversityinterp", "pctblkpopinterp",
             "pctasianpopinterp", "pctlatinpopinterp", "medincinterp",
             "pctlocalgovworker_100", "pctrentersinterp", "pctover65",
             "pctcollegegradinterp", "northeast", "south", "midwest",
             "y5", "y6", "y7", "y8", "y9")
cl <- NULL
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.9265 0.8648 -1.0713 -2.6214  0.7685  0.284
## Boot.c   -0.9265 0.8922 -1.0384 -2.7124  0.5450  0.466
## Boot.t   -0.9265 0.8648 -1.0713 -7.7805  5.9276  0.492
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -2.6757 1.6174 -1.6543 -5.8458  0.4944  0.0981
## Boot.c   -2.6757 1.7177 -1.5577 -5.6716  0.7612  0.2120
## Boot.t   -2.6757 1.6174 -1.6543 -16.6046 11.2532  0.3000
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##      2.3548  2.0000 21142.0000  0.0949
##
## $AR$ci.print
## [1] "[-6.3310, 0.3650]"
##
## $AR$ci
## [1] -6.331  0.365
##
## $AR$bounded
```

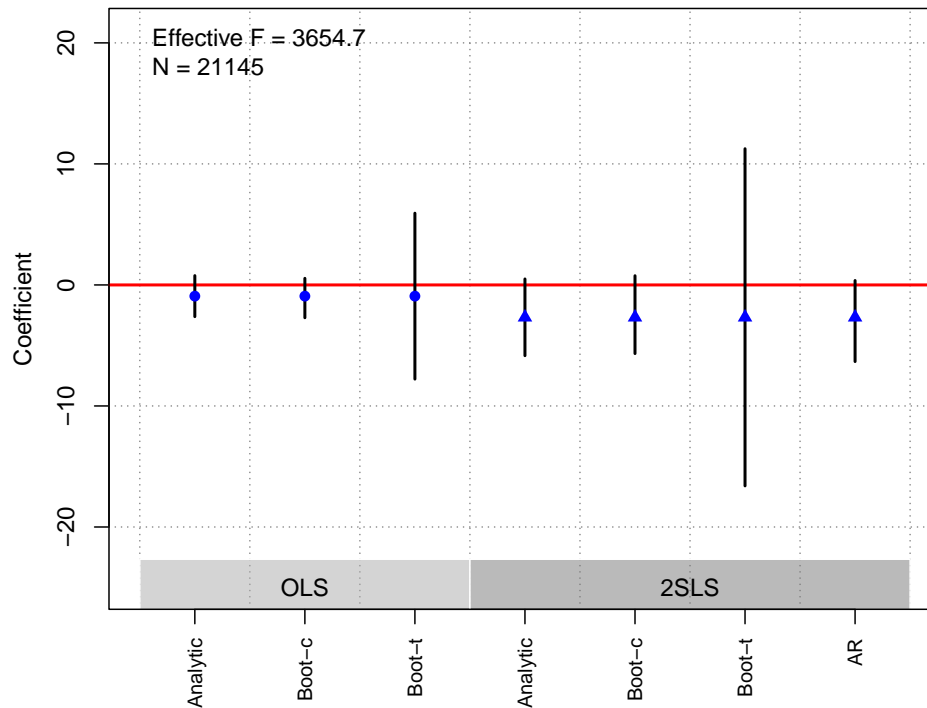
```

## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 3883.651 2506.495 NA 2513.902 3654.705
##
## $rho
## [1] 0.5185
##
## $est_rf
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## total_rivs_all -0.0081 0.0229 0.7217 0.0241 -0.0602 0.0269 0.824
## logpop -0.0855 0.0407 0.0355 0.0432 -0.1631 0.0098 0.108
##
## $est_fs
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## total_rivs_all 0.0054 3e-04 0 3e-04 0.0048 0.0060 0
## logpop 0.0291 5e-04 0 5e-04 0.0281 0.0302 0
##
## $p_iv
## [1] 2
##
## $N
## [1] 21145
##
## $N_cl
## NULL
##
## $df
## [1] 21125
##
## $nvalues
## dgepercap_cpi H_citytract_NHW_i total_rivs_all logpop
## [1,] 21129 15395 22 16223
##
## attr("class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



Vernby (2013)

Replication Summary

Unit of analysis	municipality*term
Treatment	share of noncitizens in the electorate
Instrument	immigration Inflow 1940–1950; Immigration Inflow 1960–1967
Outcome	municipal education and social spending
Model	Table3(2)

```
df<-readRDS("./rawdata/ajps_Vernby_2013.rds")
D <- "noncitvotsh"
Y <- "Y"
Z <- c("inv1950", "inv1967")
controls <- c("Taxbase2", "L_Taxbase2", "manu", "L_manu", "pop", "L_pop")
cl <- "lan"
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 8.9328 1.9684 4.5382 5.0748 12.7908 0.000
## Boot.c   8.9328 2.3503 3.8006 3.5352 12.5119 0.000
```

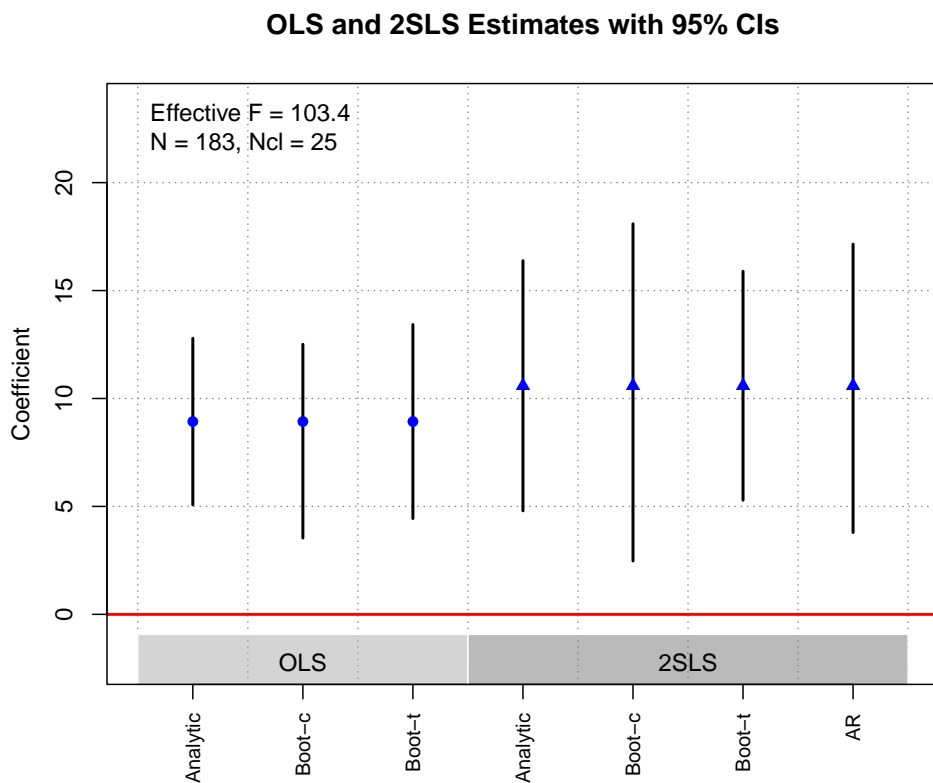
```

## Boot.t    8.9328 1.9684 4.5382  4.4406 13.4251  0.002
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 10.5903 2.9560 3.5827  4.7965 16.3840  0.0003
## Boot.c   10.5903 4.1088 2.5775  2.4722 18.0923  0.0240
## Boot.t   10.5903 2.9560 3.5827  5.2882 15.8924  0.0010
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
##   5.7276   2.0000 180.0000  0.0039
##
## $AR$ci.print
## [1] "[3.7915, 17.1525]"
##
## $AR$ci
## [1] 3.7915 17.1525
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
##   66.2203    49.5670    79.6400    25.5803    103.3586
##
## $rho
## [1] 0.6574
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## inv1950  2.5029 9.0396  0.7819 12.4069 -22.7997  25.8299  0.850
## inv1967 10.0729 7.2288  0.1635  9.5940 -9.2488  26.8433  0.244
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## inv1950  0.7234 0.3444  0.0357 0.4241 -0.1254  1.5323  0.088
## inv1967  0.4665 0.2984  0.1180 0.3394 -0.3213  0.9734  0.204
##
## $p_iv
## [1] 2
##
## $N
## [1] 183
##
## $N_c1

```

```
## [1] 25
##
## $df
## [1] 175
##
## $nvalues
##      Y noncitvotsh inv1950 inv1967
## [1,] 183      183      25      25
##
## attr(,"class")
## [1] "ivDiag"
```

plot_coef(g)



Wood and Grose (2022)

Replication Summary

Unit of analysis	House member/district
Treatment	incumbent found to have campaign finance violations
Instrument	audit
Outcome	legislator Retired
Model	Table2(1)

```
df <-readRDS("./rawdata/ajps_Wood_grose_2022.rds")
## preprocess to generate xwhat and xhat in Stata
D<-"findings"
Y <- "retire__or_resign"
Z <- "audited"
controls <-c("xwhat","south")
cl <- "stcd"
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.2369 0.1076 2.2022 0.0261 0.4477 0.0276
## Boot.c   0.2369 0.1094 2.1657 0.0342 0.4589 0.0120
## Boot.t   0.2369 0.1076 2.2022 0.0083 0.4655 0.0490
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.2869 0.1615 1.7764 -0.0297 0.6035 0.0757
## Boot.c   0.2869 0.1749 1.6407 -0.0171 0.6912 0.0660
## Boot.t   0.2869 0.1615 1.7764 -0.0432 0.6170 0.0700
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
## 2.8595 1.0000 433.0000 0.0916
##
## $AR$ci.print
## [1] "[-0.0523, 0.6390]"
##
## $AR$ci
## [1] -0.0523 0.6390
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 220.6007 22.8647 22.8647 21.1080 22.8647
##
## $rho
## [1] 0.5819
##
## $tF
```

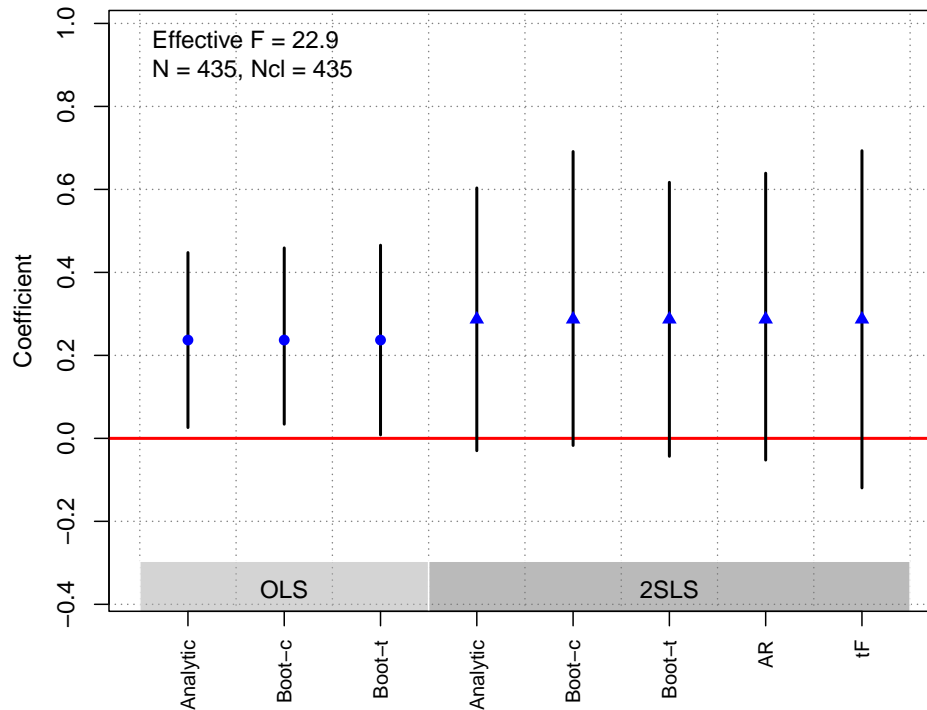
```

##      F      cF      Coef      SE      t  CI2.5% CI97.5% p-value
## 22.8647 2.5155 0.2869 0.1615 1.7764 -0.1194 0.6932 0.1663
##
## $est_rf
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## audited 0.1377 0.0816 0.0916 0.0817 -0.0066 0.307 0.066
##
## $est_fs
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## audited 0.48 0.1004 0 0.1045 0.28 0.6818 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 435
##
## $N_cl
## [1] 435
##
## $df
## [1] 431
##
## $nvalues
##      retire__or_resign findings audited
## [1,]                2          2          2
##
## attr(,"class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



Zhu (2017)

Replication Summary

Unit of analysis	province*period
Treatment	MNC activity
Instrument	weighted geographic closeness
Outcome	corruption
Model	Table1(1)

```
df <- readRDS("./rawdata/ajps_Zhu_2017.rds")
D <- "MNC"
Y <- "corruption1"
Z <- "lwdist"
controls <- c("lgdpcap6978", "gdp6978", "population", "lgovtexp9302",
             "pubempratio", "leduc", "pwratio", "female", "time")
cl <- NULL
FE <- NULL
weights <- NULL
(g <- ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
             cl =cl, weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.3531 0.0960 3.6788 0.1650 0.5412 2e-04
```

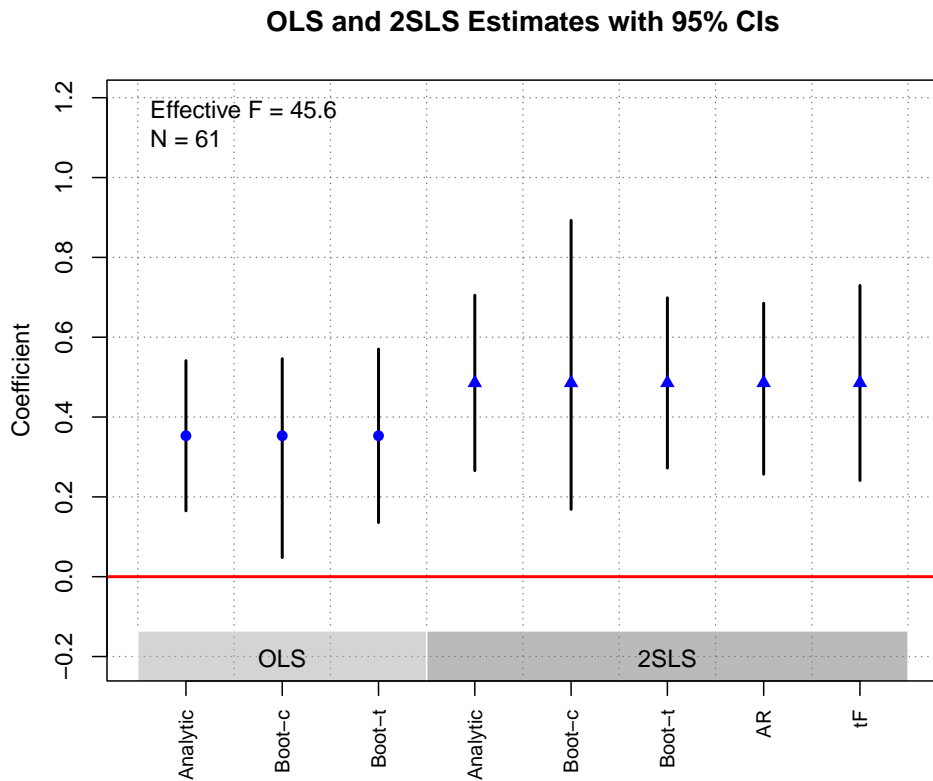
```

## Boot.c  0.3531 0.1237 2.8535  0.0480  0.5461  2e-02
## Boot.t  0.3531 0.0960 3.6788  0.1358  0.5703  4e-03
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.4855 0.1121 4.3317  0.2658  0.7052  0.000
## Boot.c   0.4855 0.2067 2.3482  0.1688  0.8927  0.006
## Boot.t   0.4855 0.1121 4.3317  0.2722  0.6987  0.001
##
## $AR
## $AR$Fstat
##      F      df1      df2      p
## 12.7838  1.0000 59.0000  0.0007
##
## $AR$ci.print
## [1] "[0.2568, 0.6850]"
##
## $AR$ci
## [1] 0.2568 0.6850
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
## 45.9155     45.5515          NA     23.8884     45.5515
##
## $rho
## [1] 0.6919
##
## $tF
##      F      cF      Coef      SE      t CI2.5% CI97.5% p-value
## 45.5515  2.1802  0.4855  0.1121  4.3317  0.2411  0.7298  0.0001
##
## $est_rf
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## lwdist 0.559 0.1698  0.001 0.2622  0.1651  1.2252  0.006
##
## $est_fs
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## lwdist 1.1514 0.1706  0 0.2356  0.8039  1.7455  0
##
## $p_iv
## [1] 1
##
## $N

```

```
## [1] 61
##
## $N_c1
## NULL
##
## $df
## [1] 50
##
## $nvalues
##      corruption1 MNC lwdist
## [1,]           61  61    61
##
## attr("class")
## [1] "ivDiag"
```

plot_coef(g)



JOP

Acharya et al. (2016)

Replication Summary

Unit of analysis county

Replication Summary

Treatment	slave proportion in 1860
Instrument	measures of the environmental suitability for growing cotton
Outcome	proportion Democrat
Model	Table2(2)

```
df<-readRDS("./rawdata/jop_Acharya_etal_2016.rds")
Y <- "dem"
D <- "pslave1860"
Z <- "cottonsuit"
controls <- c("x2", "rugged", "latitude", "x2", "longitude", "x3", "x4", "water1860")
cl <- NULL
FE <- 'code'
weights<-"sample.size"
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.0318 0.0474 -0.6701 -0.1247  0.0612  0.5028
## Boot.c   -0.0318 0.0486 -0.6534 -0.1254  0.0646  0.5120
## Boot.t   -0.0318 0.0474 -0.6701 -0.1394  0.0759  0.5540
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.2766 0.1343 -2.0596 -0.5399 -0.0134  0.0394
## Boot.c   -0.2766 0.1411 -1.9603 -0.5737 -0.0100  0.0420
## Boot.t   -0.2766 0.1343 -2.0596 -0.5462 -0.0070  0.0470
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    4.8310    1.0000 1118.0000  0.0282
##
## $AR$ci.print
## [1] "[-0.5829, -0.0322]"
##
## $AR$ci
## [1] -0.5829 -0.0322
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
```

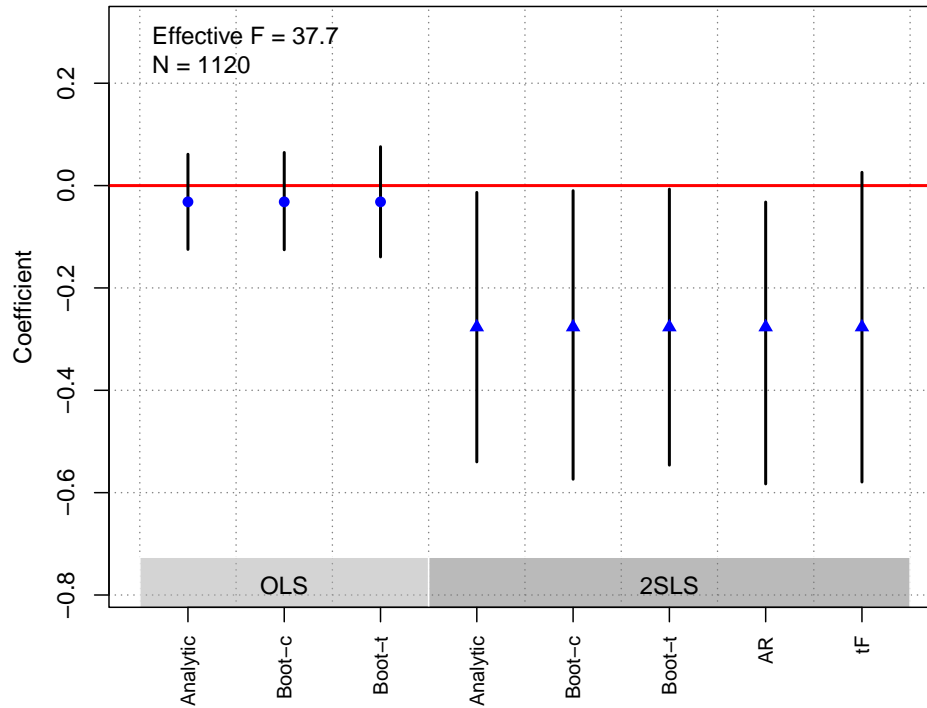
```

##      106.4957      37.6527          NA      34.6649      37.6527
##
## $rho
## [1] 0.2973
##
## $tF
##      F      cF      Coef      SE      t      CI2.5% CI97.5% p-value
## 37.6527 2.2528 -0.2766 0.1343 -2.0596 -0.5792 0.0259 0.0731
##
## $est_rf
##           Coef      SE p.value      SE.b CI.b2.5% CI.b97.5% p.value.b
## cottonsuit -0.1128 0.0518 0.0294 0.0534 -0.2161 -0.0048 0.042
##
## $est_fs
##           Coef      SE p.value      SE.b CI.b2.5% CI.b97.5% p.value.b
## cottonsuit 0.4079 0.0665 0.0000 0.0693 0.2761 0.546 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 1120
##
## $N_c1
## NULL
##
## $df
## [1] 1098
##
## $nvalues
##      dem pslave1860 cottonsuit
## [1,] 911      1077      1120
##
## attr("class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



Alt et al. (2016)

Replication Summary

Unit of analysis	individual
Treatment	unemployment expectations
Instrument	assignment to receiving an aggregate unemployment forecast
Outcome	vote intention
Model	Table2(1)

```
df<- readRDS("./rawdata/jop_Alt_etal_2015.rds")
D <- "urate_fut"
Y <- "gov"
Z <- "treatment"
controls <- "urate_now"
cl <- NULL
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.0131 0.0026 -5.0845 -0.0182 -0.0081      0
```

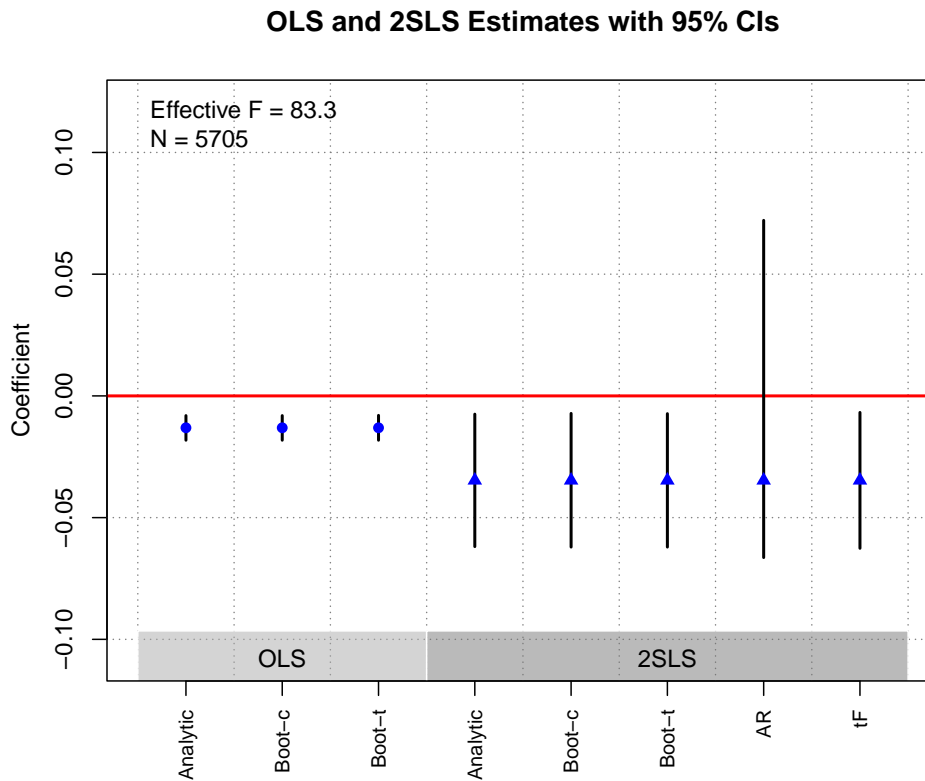
```

## Boot.c  -0.0131 0.0026 -5.0392 -0.0182 -0.0081      0
## Boot.t  -0.0131 0.0026 -5.0845 -0.0182 -0.0080      0
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.0347 0.0139 -2.5022 -0.0619 -0.0075 0.0123
## Boot.c   -0.0347 0.0141 -2.4697 -0.0621 -0.0072 0.0040
## Boot.t   -0.0347 0.0139 -2.5022 -0.0621 -0.0073 0.0090
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    0.0017    1.0000 5703.0000    0.9671
##
## $AR$ci.print
## [1] "[-0.0664, 0.0721]"
##
## $AR$ci
## [1] -0.0664 0.0721
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
##    60.1863    68.9098         NA    67.1063    83.3152
##
## $rho
## [1] 0.0801
##
## $tF
##           F      cF      Coef      SE      t CI2.5% CI97.5% p-value
##    83.3152  2.0100 -0.0347  0.0139 -2.5022 -0.0626 -0.0068 0.0147
##
## $est_rf
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## treatment 0.027 0.0243 0.2661 0.0238 -0.0197 0.0743 0.22
##
## $est_fs
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## treatment -0.9354 0.1169 0 0.1142 -1.1621 -0.7159 0
##
## $p_iv
## [1] 1
##
## $N

```

```
## [1] 5705
##
## $N_c1
## NULL
##
## $df
## [1] 5702
##
## $nvalues
##      gov urate_fut treatment
## [1,] 2      88      8
##
## attr("class")
## [1] "ivDiag"
```

`plot_coef(g)`



Arias and Stasavage (2019)

Replication Summary

Unit of analysis	country*year
Treatment	government expenditures
Instrument	trade shock × UK bond yield
Outcome	regular leader turnover

Replication Summary

Model Table3(2)

```
# Variables are already residualized against controls, fixed effects, and unit-specific trends
df<-readRDS("./rawdata/jop_Arias_etal_2019.rds")
Y <- "regular_res"
D <- "d expenditures_res"
Z <- "interact_res"
controls <- NULL
cl<-c("cocode", "year")
FE<-NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.0215 0.0359 -0.5975 -0.0919  0.0490  0.5502
## Boot.c   -0.0215 0.0391 -0.5499 -0.0910  0.0612  0.6053
## Boot.t   -0.0215 0.0359 -0.5975 -0.0744  0.0315  0.4656
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic  0.8282  1.6891  0.4903 -2.4824  4.1389  0.6239
## Boot.c    0.8282 20.8111 0.0398 -1.6956  9.7722  0.5101
## Boot.t    0.8282  1.6891  0.4903 -1.5755  3.2320  0.4370
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    0.2643    1.0000 2743.0000  0.6073
##
## $AR$ci.print
## [1] "[-2.1784, 5.7604]"
##
## $AR$ci
## [1] -2.1784  5.7604
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
##    3.0429    3.4739   14.4763    7.5067   14.4763
##
```

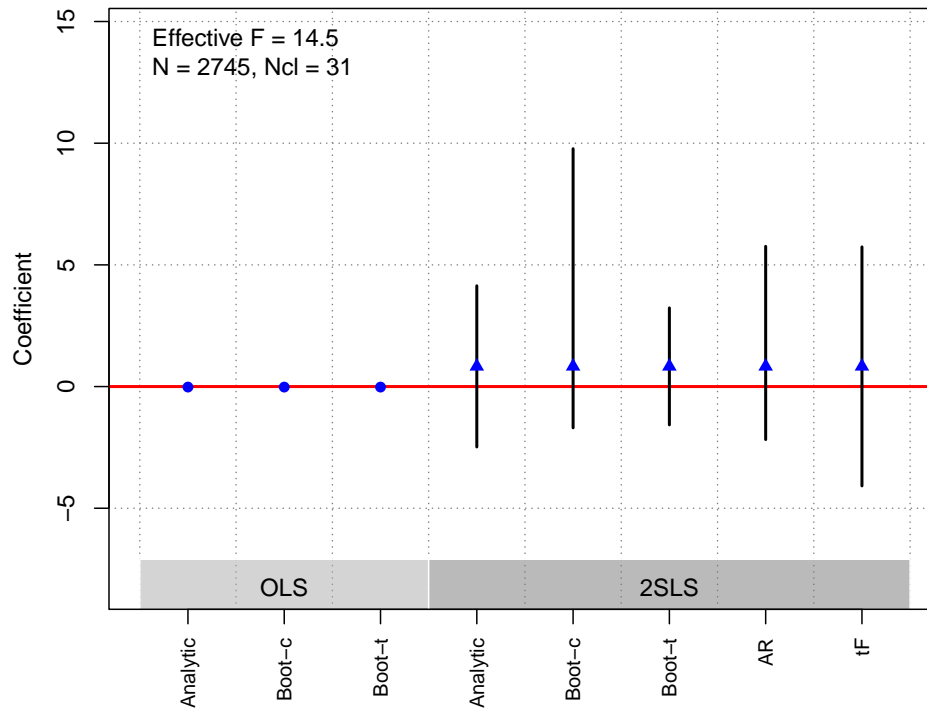
```

## $rho
## [1] 0.0333
##
## $tF
##      F      cF      Coef      SE      t  CI2.5% CI97.5% p-value
## 14.4763  2.9071  0.8282  1.6891  0.4903 -4.0822  5.7387  0.7410
##
## $est_rf
##              Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## interact_res 0.276 0.5369  0.6072 0.4693 -0.4889  1.3784  0.4762
##
## $est_fs
##              Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## interact_res 0.3332 0.0876  1e-04 0.1216  0.0477  0.5343  0.0339
##
## $p_iv
## [1] 1
##
## $N
## [1] 2745
##
## $N_cl
## [1] 31
##
## $df
## [1] 2743
##
## $nvalues
##      regular_res d expenditures_res interact_res
## [1,]          2745          2745          2745
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



Bhavnani and Lee (2018)

Replication Summary

Unit of analysis	district*period
Treatment	bureaucrats' embeddedness
Instrument	early-career job assignment
Outcome	proportion of villages with high schools
Model	Table1(4)

```
df <- readRDS("./rawdata/jop_Bhavnani_etal_2018.rds")
D <- "ALLlocal"
Y <- "Phigh"
Z <- "EXALLlocal"
controls <- c("ALLbachdivi", "lnnewpop", "lnnvill", "p_rural", "p_work",
             "p_aglab", "p_sc", "p_st", "lnmurderpc", "stategov", "natgov")
cl <- "distcode71"
FE <- c('distcode71', "year")
weights <- NULL
(g <- ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
             cl =cl, weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0195 0.0073 2.6753 0.0052 0.0337 0.0075
```

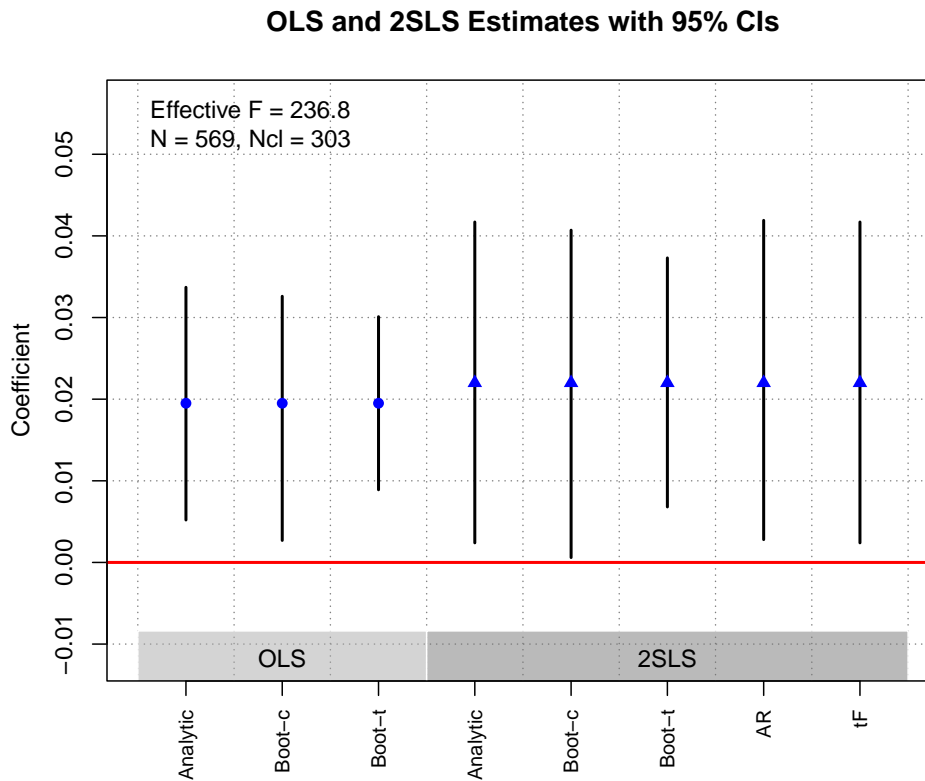
```

## Boot.c  0.0195 0.0077 2.5363  0.0027  0.0326  0.0200
## Boot.t  0.0195 0.0073 2.6753  0.0089  0.0301  0.0000
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.022 0.0100 2.1986  0.0024  0.0417  0.0279
## Boot.c   0.022 0.0103 2.1365  0.0006  0.0407  0.0340
## Boot.t   0.022 0.0100 2.1986  0.0068  0.0373  0.0030
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
##  5.0041  1.0000 567.0000  0.0257
##
## $AR$ci.print
## [1] "[0.0028, 0.0419]"
##
## $AR$ci
## [1] 0.0028 0.0419
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
##  243.2947  215.8574  236.8206  240.8167  236.8206
##
## $rho
## [1] 0.7002
##
## $tF
##          F      cF      Coef      SE      t  CI2.5%  CI97.5%  p-value
## 236.8206  1.9600  0.0220  0.0100  2.1986  0.0024  0.0417  0.0279
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## EXALLlocal 0.0121 0.0055 0.0267 0.0057  3e-04  0.0224  0.034
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## EXALLlocal 0.5504 0.0358  0 0.0355  0.4842  0.616  0
##
## $p_iv
## [1] 1
##
## $N

```

```
## [1] 569
##
## $N_cl
## [1] 303
##
## $df
## [1] 253
##
## $nvalues
##      Phigh ALLlocal EXALLlocal
## [1,] 567      493      318
##
## attr("class")
## [1] "ivDiag"
```

`plot_coef(g)`



Charron and Lapuente (2013)

Replication Summary

Unit of analysis
 Treatment
 Instrument

region
 clientelism
 consolidation of clientelistic networks in regions where rulers have historically less constraints to their decisions

Replication Summary

Outcome quality of governments
Model Table3(2a)

```
df<-readRDS("./rawdata/jop_Charron_etal_2013.rds")
D <- "pc_all4_tol"
Y <- "eqi"
Z <- c("pc_institutions","literacy1880")
controls <- c("logpop", "capitalregion", "ger", "it", "uk","urb_1860_1850_30")
cl <- NULL
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0176 0.0034 5.186 0.0110 0.0243 0
## Boot.c   0.0176 0.0034 5.130 0.0104 0.0239 0
## Boot.t   0.0176 0.0034 5.186 0.0103 0.0250 0
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0233 0.0041 5.7196 0.0153 0.0313 0
## Boot.c   0.0233 0.0042 5.5429 0.0146 0.0312 0
## Boot.t   0.0233 0.0041 5.7196 0.0148 0.0318 0
##
## $AR
## $AR$Fstat
##      F      df1      df2      p
## 18.2062 2.0000 53.0000 0.0000
##
## $AR$ci.print
## [1] "[0.0170, 0.0297]"
##
## $AR$ci
## [1] 0.0170 0.0297
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 37.2005 31.2712 NA 30.0615 19.9514
##
```

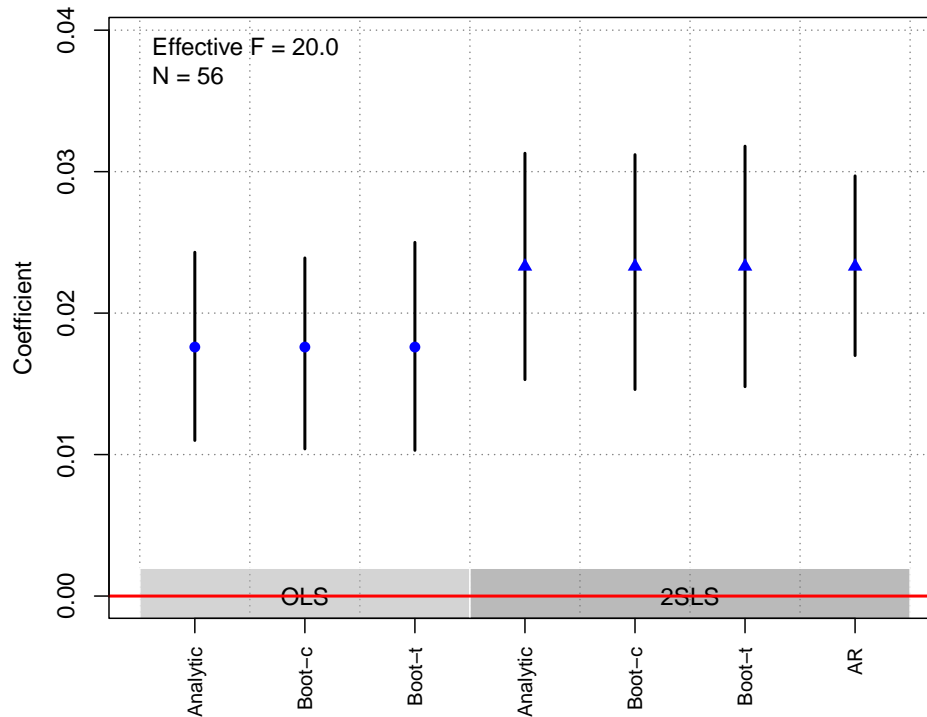
```

## $rho
## [1] 0.7828
##
## $est_rf
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## pc_institutions 0.1941 0.0765  0.0111 0.0791  0.0408  0.354  0.010
## literacy1880    0.0204 0.0043  0.0000 0.0050  0.0092  0.029  0.002
##
## $est_fs
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## pc_institutions 12.1093 2.3469  0e+00 2.4409  7.4604 16.8320  0.000
## literacy1880    0.5348 0.1319  1e-04 0.1499  0.1920  0.7925  0.006
##
## $p_iv
## [1] 2
##
## $N
## [1] 56
##
## $N_c1
## NULL
##
## $df
## [1] 48
##
## $nvalues
##      eqi pc_all4_tol pc_institutions literacy1880
## [1,] 56          44              14              38
##
## attr("class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



Charron et al. (2017)

Replication Summary

Unit of analysis	region
Treatment	more developed bureaucracy
Instrument	proportion of Protestant residents in a region; aggregate literacy in 1880
Outcome	percent of single bidders in procurement contracts
Model	Table5(4)

```
df <- readRDS("./rawdata/jop_Charron_etal_2017.rds")
D <- "pubmerit"
Y <- "lcri_euc1_r"
Z <- c("lirate_1880", 'pctprot')
controls <- c("logpopdens", "logppp11", "trust", "pctwomenpar1")
cl <- "country"
FE <- NULL
weights<-"eu_popweights"
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.09 0.0155 -5.8068 -0.1204 -0.0597 0.000
## Boot.c   -0.09 0.0233 -3.8649 -0.1105 -0.0206 0.004
```

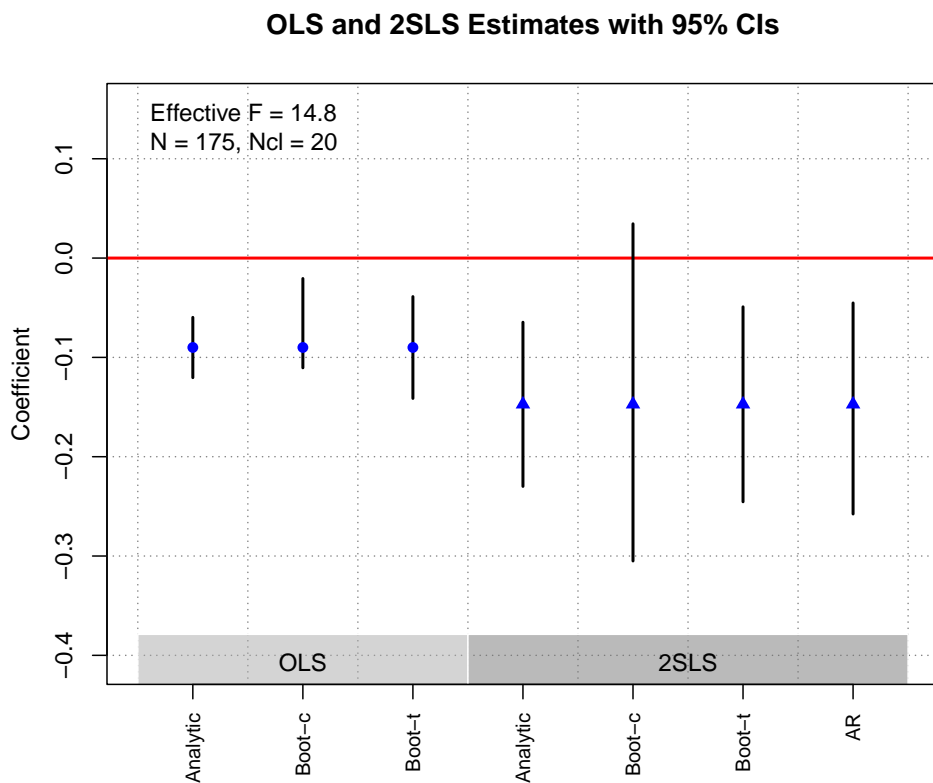
```

## Boot.t   -0.09 0.0155 -5.8068 -0.1413  -0.0388   0.003
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.1472 0.0422 -3.4919 -0.2299  -0.0646  0.0005
## Boot.c   -0.1472 0.1001 -1.4712 -0.3050   0.0345  0.0900
## Boot.t   -0.1472 0.0422 -3.4919 -0.2454  -0.0491  0.0120
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##   5.5325  2.0000 172.0000  0.0047
##
## $AR$ci.print
## [1] "[-0.2577, -0.0452]"
##
## $AR$ci
## [1] -0.2577 -0.0452
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
##   27.8775    23.2292    36.2651     6.2999    14.8219
##
## $rho
## [1] 0.4992
##
## $est_rf
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## litrate_1880 -0.0009 0.0005  0.0767 0.0006  -0.0019   0.0004   0.184
## pctprot      -0.1769 0.1131  0.1177 0.1417  -0.4289   0.1099   0.288
##
## $est_fs
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## litrate_1880 0.0060 0.0025  0.0184 0.0030   0.0000   0.0115   0.050
## pctprot      1.1959 0.3235  0.0002 0.4935   0.0099   1.9314   0.048
##
## $p_iv
## [1] 2
##
## $N
## [1] 175
##
## $N_c1

```

```
## [1] 20
##
## $df
## [1] 169
##
## $nvalues
##      lcri_euc1_r pubmerit litrate_1880 pctprot
## [1,]          173          173           78      131
##
## attr(,"class")
## [1] "ivDiag"
```

plot_coef(g)



Cirone and Van Coppenolle (2018)

Replication Summary

Unit of analysis	deputy*year
Treatment	budget committee service
Instrument	random assignment of budget incumbents to bureaux
Outcome	legislator sponsorship on a budget bill
Model	Table2(2)

```

df<- readRDS("./rawdata/jop_Cirone_etal_2018.rds")
D <- "budget"
Y <- "F1to5billbudgetdummy"
Z <- "bureauotherbudgetincumbent"
controls <- c("budgetincumbent", "cummyears", "cummyears2",
              "age", "age2", "permargin", "permargin2",
              "inscrits", "inscrits2", "proprietaire",
              "lib_all", "civil", "paris")
cl <- c("id", "year")
FE <- "year"
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))

## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0305 0.0218 1.3957 -0.0123  0.0733  0.1628
## Boot.c   0.0305 0.0181 1.6792 -0.0042  0.0662  0.0940
## Boot.t   0.0305 0.0218 1.3957 -0.0003  0.0612  0.0530
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.6341 0.3548 1.7872 -0.0613  1.3295  0.0739
## Boot.c   0.6341 0.2660 2.3840  0.1800  1.2455  0.0100
## Boot.t   0.6341 0.3548 1.7872  0.1965  1.0717  0.0120
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    3.0669    1.0000 8145.0000  0.0799
##
## $AR$ci.print
## [1] "[-0.0755, 1.3224]"
##
## $AR$ci
## [1] -0.0755  1.3224
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
##    32.1302    34.2557   168.0023    33.4141    168.0023
##
## $rho
## [1] 0.0628

```

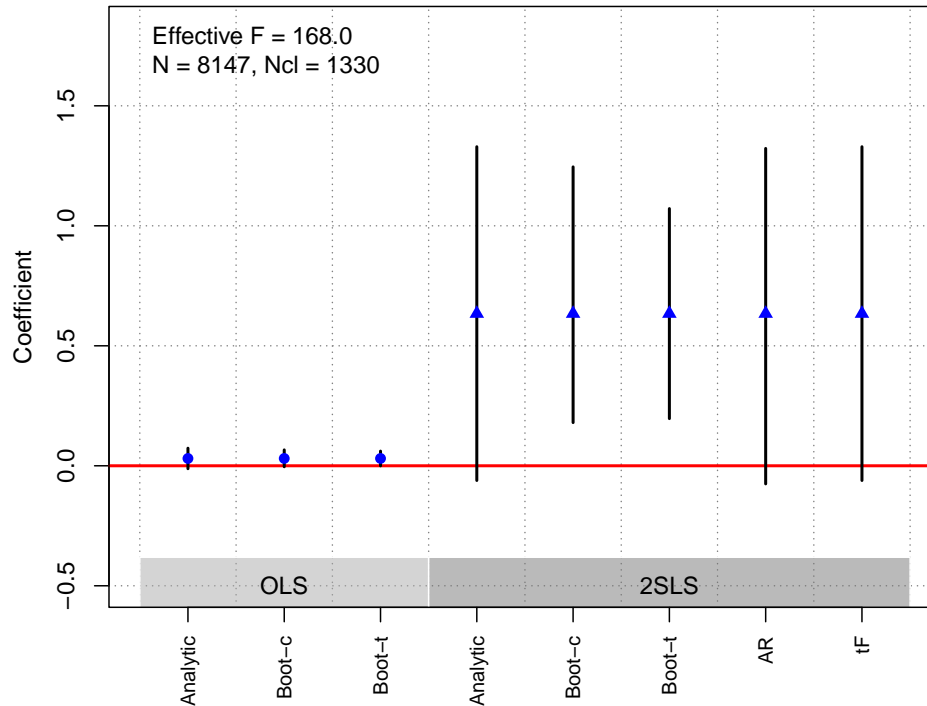
```

##
## $tF
##      F      cF      Coef      SE      t      CI2.5% CI97.5% p-value
## 168.0023  1.9600  0.6341  0.3548  1.7872  -0.0613  1.3295  0.0739
##
## $est_rf
##
##      Coef      SE p.value      SE.b CI.b2.5% CI.b97.5%
## bureauotherbudgetincumbent -0.0052 0.003  0.0801  0.0018  -0.0088  -0.0016
##
##      p.value.b
## bureauotherbudgetincumbent      0.01
##
## $est_fs
##
##      Coef      SE p.value      SE.b CI.b2.5% CI.b97.5%
## bureauotherbudgetincumbent -0.0083 6e-04      0 0.0014  -0.0113  -0.0054
##
##      p.value.b
## bureauotherbudgetincumbent      0
##
## $p_iv
## [1] 1
##
## $N
## [1] 8147
##
## $N_c1
## [1] 1330
##
## $df
## [1] 13
##
## $nvalues
##      F1to5billbudgetdummy budget bureauotherbudgetincumbent
## [1,]                      2      2                          9
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



Dietrich and Wright (2015)

Replication Summary

Unit of analysis	transition
Treatment	economic aid
Instrument	constructed Z
Outcome	transitions to multipartyism
Model	Table1(2)

```
df <- readRDS("./rawdata/jop_Dietrich_2015.rds")
D <- "econaid"
Y <- "mp"
Z <- c("linfl3", "econaid_lgdp_g", "econaid_lpop_g",
      "econaid_cwar_g", "econaid_dnmp_g",
      "econaid_dnmp2_g", "econaid_dnmp3_g")
controls <- c('lgdp', 'lpop', 'cwar', 'dmp',
             'dmp2', 'dmp3', 'dnmp', 'dnmp2', 'dnmp3')
cl <- "cowcode"
FE <- NULL
weights <- NULL
(g <- ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
             cl =cl, weights=weights, cores = cores))
```

```
## $est_ols
```

```

##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0576 0.0233 2.4734 0.0119 0.1032 0.0134
## Boot.c   0.0576 0.0288 1.9971 -0.0117 0.1030 0.1020
## Boot.t   0.0576 0.0233 2.4734 0.0226 0.0925 0.0020
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.1075 0.0401 2.6795 0.0289 0.1861 0.0074
## Boot.c   0.1075 0.0505 2.1292 -0.0066 0.2001 0.0700
## Boot.t   0.1075 0.0401 2.6795 0.0337 0.1813 0.0030
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
## 3.5039 7.0000 362.0000 0.0012
##
## $AR$ci.print
## [1] "[0.0361, 0.2102]"
##
## $AR$ci
## [1] 0.0361 0.2102
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 28.9900 47.6878 22.5931 1.8693 5.4068
##
## $rho
## [1] 0.6026
##
## $est_rf
##           Coef      SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## Iinfl3      0.0382 0.0180 0.0341 0.0229 -0.0138 0.0774 0.160
## econaid_lgdp_g 0.0459 0.0246 0.0624 0.0504 0.0047 0.2038 0.034
## econaid_lpop_g 0.0049 0.0218 0.8229 0.0356 -0.0450 0.0987 0.746
## econaid_cwar_g -0.0084 0.0635 0.8946 0.1004 -0.2305 0.1684 0.930
## econaid_dnmp_g -0.0227 0.0268 0.3965 0.0328 -0.0763 0.0481 0.592
## econaid_dnmp2_g 0.0010 0.0011 0.3704 0.0015 -0.0023 0.0034 0.652
## econaid_dnmp3_g 0.0000 0.0000 0.4243 0.0000 0.0000 0.0000 0.772
##
## $est_fs
##           Coef      SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## Iinfl3      0.1561 0.0506 0.0020 0.0620 0.0044 0.2379 0.044
## econaid_lgdp_g 0.1664 0.1524 0.2749 0.2697 -0.5115 0.6424 0.596

```

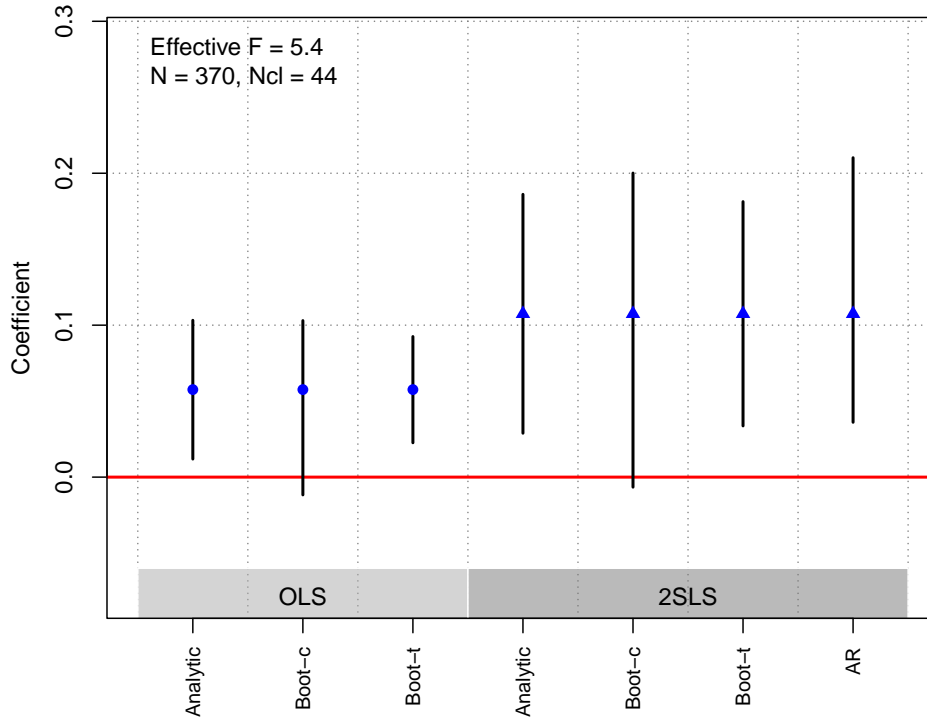
```

## econaid_lpop_g    0.1839 0.0976  0.0596 0.1664 -0.3098    0.3684    0.388
## econaid_cwar_g   -0.2848 0.3413  0.4041 0.5017 -1.6604    0.3719    0.462
## econaid_dnmp_g   -0.0235 0.0899  0.7933 0.1039 -0.2589    0.1326    0.734
## econaid_dnmp2_g  -0.0009 0.0045  0.8455 0.0054 -0.0087    0.0119    0.998
## econaid_dnmp3_g  0.0000 0.0001  0.5707 0.0001 -0.0001    0.0001    0.852
##
## $p_iv
## [1] 7
##
## $N
## [1] 370
##
## $N_cl
## [1] 44
##
## $df
## [1] 362
##
## $nvalues
##      mp econaid Iinfl3 econaid_lgdp_g econaid_lpop_g econaid_cwar_g
## [1,] 2      370      370              370              370              370
##      econaid_dnmp_g econaid_dnmp2_g econaid_dnmp3_g
## [1,]              370              370              370
##
## attr("class")
## [1] "ivDiag"

```

plot_coef(g)

OLS and 2SLS Estimates with 95% CIs



DiGiuseppe and Shea (2022)

Replication Summary

Unit of analysis	country*year
Treatment	US support
Instrument	echelon corridor
Outcome	property rights
Model	Table1(5)

```
df <-readRDS("./rawdata/jop_digiuseppe_2022.rds")
D <- "wi_usa_median"
Y<-"Fwi_v2stfiscap2"
Z <- "Echelon2"
controls <-c("wi_v2xcl_prpty","wi_compete", "wi_lnpop_wdi",
            "wi_lngdppc", "wi_polity2", "wi_polity2_2", "wi_ny_gdp_totl_rt_zs",
            "wi_cwyr", "wi_c2", "wi_c3", "coldwar")

cl<- NULL
FE<- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
```

```

## Analytic 0.0443 0.0156 2.8331 0.0136 0.0749 0.0046
## Boot.c 0.0443 0.0156 2.8338 0.0145 0.0741 0.0060
## Boot.t 0.0443 0.0156 2.8331 0.0142 0.0743 0.0050
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.8158 0.3217 2.5360 0.1853 1.4463 0.0112
## Boot.c 0.8158 0.7680 1.0623 0.2443 1.8771 0.0060
## Boot.t 0.8158 0.3217 2.5360 0.1931 1.4385 0.0070
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
##    8.5251    1.0000 2366.0000 0.0035
##
## $AR$ci.print
## [1] "[0.2818, 1.8803]"
##
## $AR$ci
## [1] 0.2818 1.8803
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
##    18.8218    12.1084         NA    12.0737    12.1084
##
## $rho
## [1] 0.089
##
## $tF
##          F      cF      Coef      SE      t CI2.5% CI97.5% p-value
## 12.1084  3.1262  0.8158  0.3217  2.5360 -0.1899  1.8215  0.1118
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## Echelon2 0.1792 0.0615 0.0036 0.0601 0.0548 0.2929 0.004
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## Echelon2 0.2196 0.0631 5e-04 0.0632 0.0864 0.3415 0.004
##
## $p_iv
## [1] 1
##

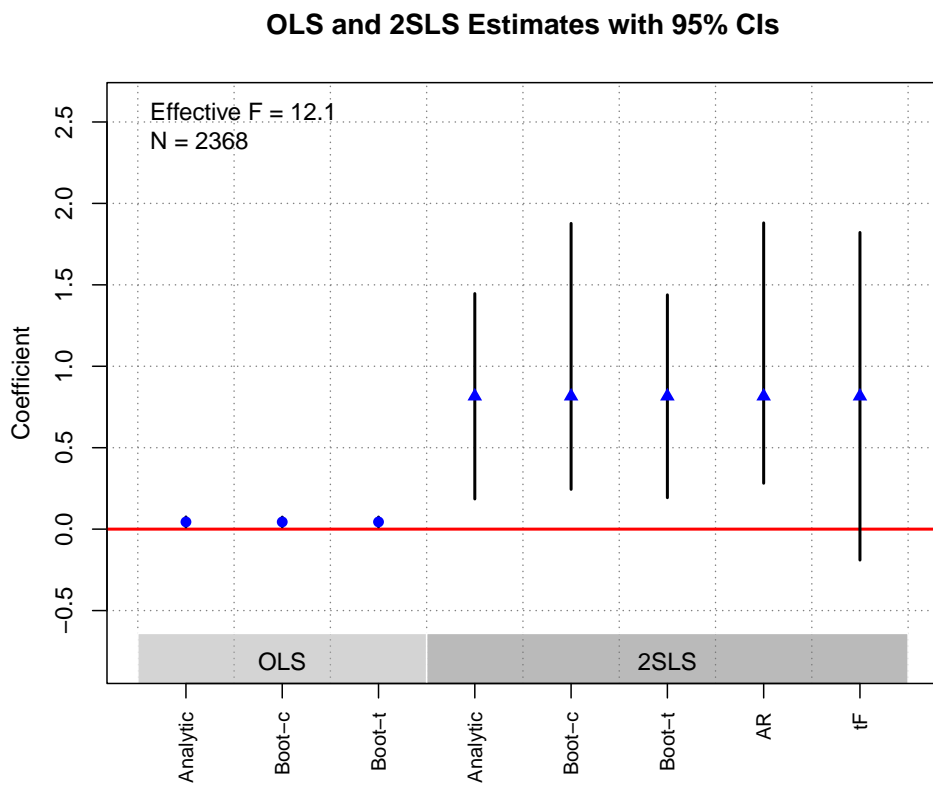
```

```

## $N
## [1] 2368
##
## $N_c1
## NULL
##
## $df
## [1] 2355
##
## $nvalues
##      Fwi_v2stfiscap2 wi_usa_median Echelon2
## [1,]                314          2368      2
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`



Dube and Naidu (2015)

Replication Summary

Unit of analysis	municipality*year
Treatment	changes in US funding to Colombia
Instrument	US funding in countries outside of Latin America

Replication Summary

Outcome the number of paramilitary attacks
Model Table1(1)

```
df<-readRDS("./rawdata/jop_Dube_etal_2015.rds")
D <- "bases6xlrmlnar_col"
Y <- "paratt"
Z <- "bases6xlrmlwnl"
controls <- "lnnewpop"
cl <- "municipality"
FE <- c("year","municipality")
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.1503 0.0601 2.5001 0.0325 0.2682 0.0124
## Boot.c   0.1503 0.0610 2.4632 0.0441 0.2770 0.0080
## Boot.t   0.1503 0.0601 2.5001 0.0402 0.2604 0.0160
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.3149 0.1212 2.5977 0.0773 0.5525 0.0094
## Boot.c   0.3149 0.1231 2.5580 0.0972 0.5678 0.0020
## Boot.t   0.3149 0.1212 2.5977 0.0866 0.5433 0.0190
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    6.7529    1.0000 16604.0000 0.0094
##
## $AR$ci.print
## [1] "[0.0797, 0.5525]"
##
## $AR$ci
## [1] 0.0797 0.5525
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
##    7003.8727    810.8395 185092.5288 176287.6694 185092.5288
##
```

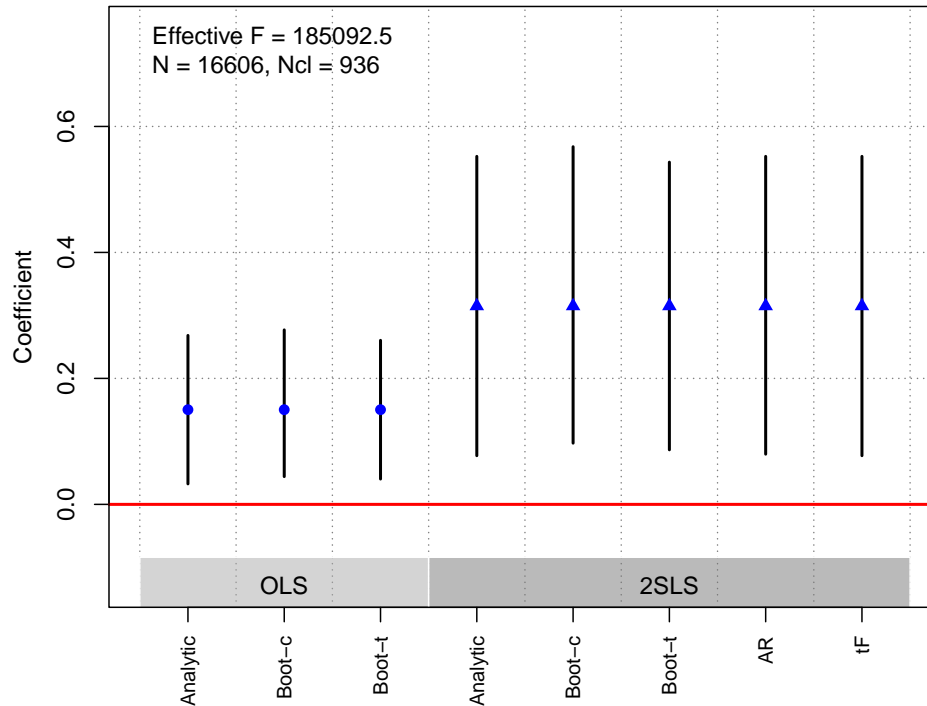
```

## $rho
## [1] 0.556
##
## $tF
##           F           cF           Coef           SE           t           CI2.5%
## 185092.5288      1.9600      0.3149      0.1212      2.5977      0.0773
##           CI97.5%      p-value
##           0.5525      0.0094
##
## $est_rf
##           Coef           SE p.value      SE.b CI.b2.5% CI.b97.5% p.value.b
## bases6xlrmlwnl 1.1155 0.4293 0.0094 0.4362 0.3434 2.0122 0.002
##
## $est_fs
##           Coef           SE p.value      SE.b CI.b2.5% CI.b97.5% p.value.b
## bases6xlrmlwnl 3.5422 0.0082 0 0.0084 3.5227 3.5562 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 16606
##
## $N_c1
## [1] 936
##
## $df
## [1] 935
##
## $nvalues
##           paratt bases6xlrmlnar_col bases6xlrmlwnl
## [1,]      13              19              18
##
## attr("class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



Feigenbaum and Hall (2015)

Replication Summary

Unit of analysis	congressional district*decade
Treatment	localized trade shocks in congressional districts
Instrument	Chinese exports to other economies*local exposure
Outcome	trade score based on congressional voting
Model	Table1(3)

```
df<-readRDS("./rawdata/jop_Feigenbaum_etal_2015.rds")
D <-"x"
Y <- "tradescore"
Z <- "z"
controls <- c("dem_share")
cl <- "state_cluster"
FE <- "decade"
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.108 0.2965 -0.3643 -0.6891  0.4731  0.7157
## Boot.c   -0.108 0.3052 -0.3539 -0.7021  0.4620  0.7160
```

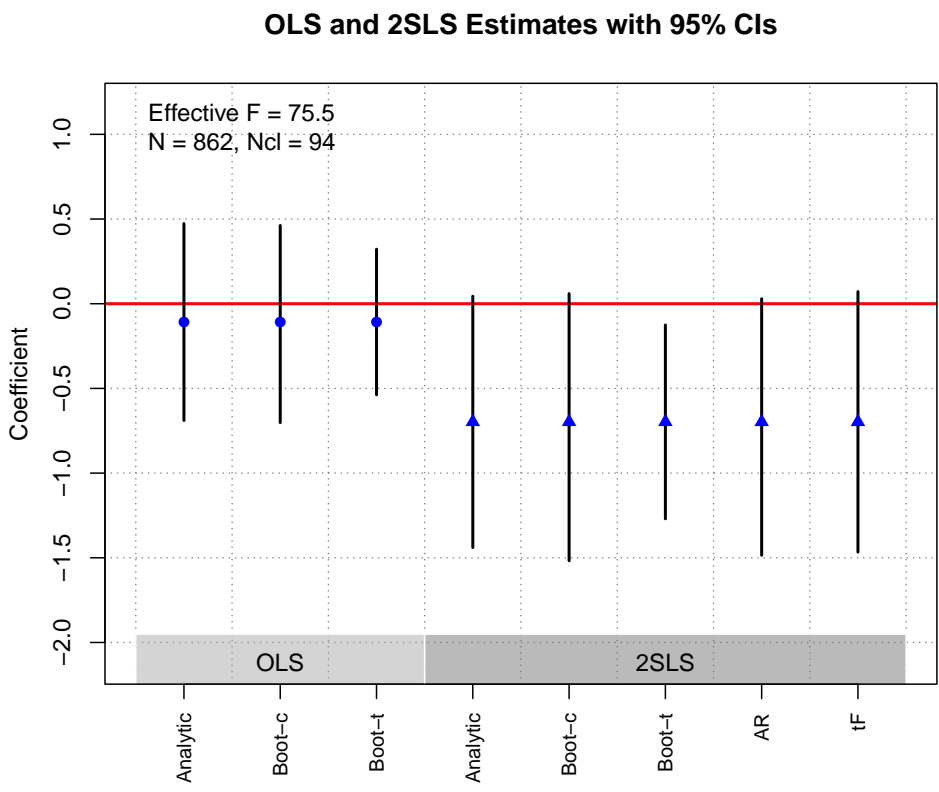
```

## Boot.t   -0.108 0.2965 -0.3643 -0.5377   0.3217  0.5980
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.6976 0.3786 -1.8424 -1.4397   0.0445  0.0654
## Boot.c   -0.6976 0.4032 -1.7301 -1.5175   0.0594  0.0680
## Boot.t   -0.6976 0.3786 -1.8424 -1.2698  -0.1254  0.0150
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##   3.4825   1.0000 860.0000  0.0624
##
## $AR$ci.print
## [1] "[-1.4852, 0.0294]"
##
## $AR$ci
## [1] -1.4852  0.0294
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
## 1189.3393   204.4798   75.5233   66.0848   75.5233
##
## $rho
## [1] 0.7622
##
## $tF
##           F      cF      Coef      SE      t CI2.5% CI97.5% p-value
## 75.5233  2.0310 -0.6976  0.3786 -1.8424 -1.4666  0.0714  0.0754
##
## $est_rf
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## z -0.5863 0.3145  0.0623 0.3485 -1.3311  0.0505  0.068
##
## $est_fs
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## z 0.8405 0.0967  0 0.1034  0.684  1.0772  0
##
## $p_iv
## [1] 1
##
## $N
## [1] 862

```

```
##
## $N_cl
## [1] 94
##
## $df
## [1] 858
##
## $nvalues
##      tradescore    x    z
## [1,]          709 698 697
##
## attr("class")
## [1] "ivDiag"
```

plot_coef(g)



Flores-Macias and Kreps (2013)

Replication Summary

Unit of analysis	country*year
Treatment	trade volume
Instrument	lagged energy production
Outcome	foreign policy convergence
Model	Table2(1)

```

df<- readRDS("./rawdata/jop_Flores_etal_2013.rds")
D <- "log_tot_trade"
Y <- "log_HRVOTE"
Z <- "lag_log_energ_prod"
controls <- c("log_cinc", "us_aid100", "log_tot_ustrate",
              "Joint_Dem_Dum", "pts_score", "dummy2004")
cl <- NULL
FE <- 'statea'
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
            cl =cl,weights=weights, cores = cores))

```

```

## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0191 0.0044 4.3531 0.0105 0.0277 0
## Boot.c   0.0191 0.0045 4.2568 0.0110 0.0284 0
## Boot.t   0.0191 0.0044 4.3531 0.0104 0.0277 0
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0456 0.0135 3.3747 0.0191 0.0721 7e-04
## Boot.c   0.0456 0.0150 3.0438 0.0182 0.0773 4e-03
## Boot.t   0.0456 0.0135 3.3747 0.0181 0.0731 3e-03
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
## 14.1713 1.0000 590.0000 0.0002
##
## $AR$ci.print
## [1] "[0.0218, 0.0745]"
##
## $AR$ci
## [1] 0.0218 0.0745
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 66.1143 53.6345 NA 50.8247 53.6345
##
## $rho
## [1] 0.3295
##
## $tF

```

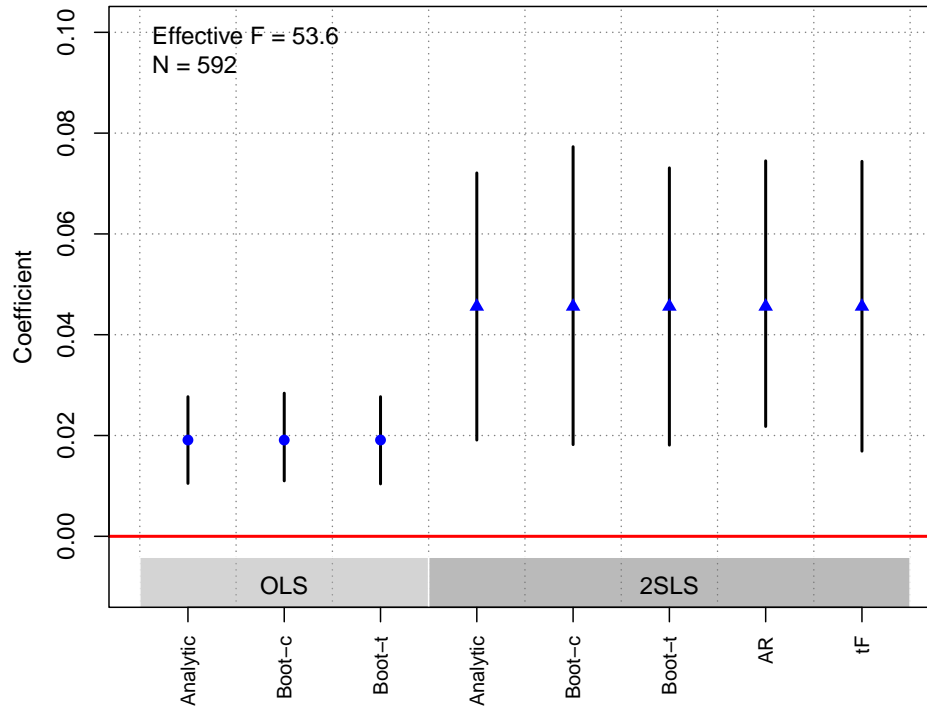
```

##      F      cF      Coef      SE      t  CI2.5% CI97.5% p-value
## 53.6345 2.1276 0.0456 0.0135 3.3747 0.0169 0.0744 0.0019
##
## $est_rf
##              Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## lag_log_energ_prod 0.1086 0.0301 3e-04 0.0324 0.0426 0.1696 0.004
##
## $est_fs
##              Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## lag_log_energ_prod 2.3803 0.325 0 0.3339 1.7522 3.0469 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 592
##
## $N_c1
## NULL
##
## $df
## [1] 543
##
## $nvalues
##      log_HRVOTE log_tot_trade lag_log_energ_prod
## [1,]          32          590          581
##
## attr(,"class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



Gehlbach and Keefer (2012)

Replication Summary

Unit of analysis	nondemocratic episode
Treatment	age of ruling party less leader years in office
Instrument	whether the first ruler in a nondemocratic episode is a military leader
Outcome	private invest
Model	Table1(4)

```
df<- readRDS("./rawdata/jop_Gelbach_etal_2012.rds")
D <- "gov1_yrs"
Y <- "gfcf_priv_gdp"
Z <- "military_first_alt"
controls <- c("tenure", "stabs", "fuelex_gdp", "oresex_gdp",
             "frac_ethn", "frac_relig", "frac_ling", "pop_yng_pct",
             "pop_tot", "pop_ru_pct", "land_km", "gdppc_ppp_2005_us")
cl <- "ifs_code"
FE <-NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
```

```

## Analytic 0.1304 0.0351 3.7118 0.0615 0.1992 2e-04
## Boot.c 0.1304 0.0424 3.0757 0.0482 0.2173 4e-03
## Boot.t 0.1304 0.0351 3.7118 0.0646 0.1961 0e+00
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.3956 0.1798 2.2001 0.0432 0.7479 0.0278
## Boot.c 0.3956 1.0389 0.3807 0.1058 1.0507 0.0140
## Boot.t 0.3956 0.1798 2.2001 0.1151 0.6760 0.0120
##
## $AR
## $AR$Fstat
##      F      df1      df2      p
## 6.3658 1.0000 97.0000 0.0133
##
## $AR$ci.print
## [1] "[0.0971, 0.9654]"
##
## $AR$ci
## [1] 0.0971 0.9654
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
##      6.3713      9.2042      9.5714      9.4400      9.5714
##
## $rho
## [1] 0.2641
##
## $tF
##      F      cF      Coef      SE      t CI2.5% CI97.5% p-value
## 9.5714 3.5187 0.3956 0.1798 2.2001 -0.2371 1.0282 0.2204
##
## $est_rf
##          Coef      SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## military_first_alt -3.3385 1.4135 0.0182 1.3984 -6.1212 -0.8649 0.01
##
## $est_fs
##          Coef      SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## military_first_alt -8.4401 2.7281 0.002 2.747 -14.1454 -3.028 0.004
##
## $p_iv
## [1] 1
##

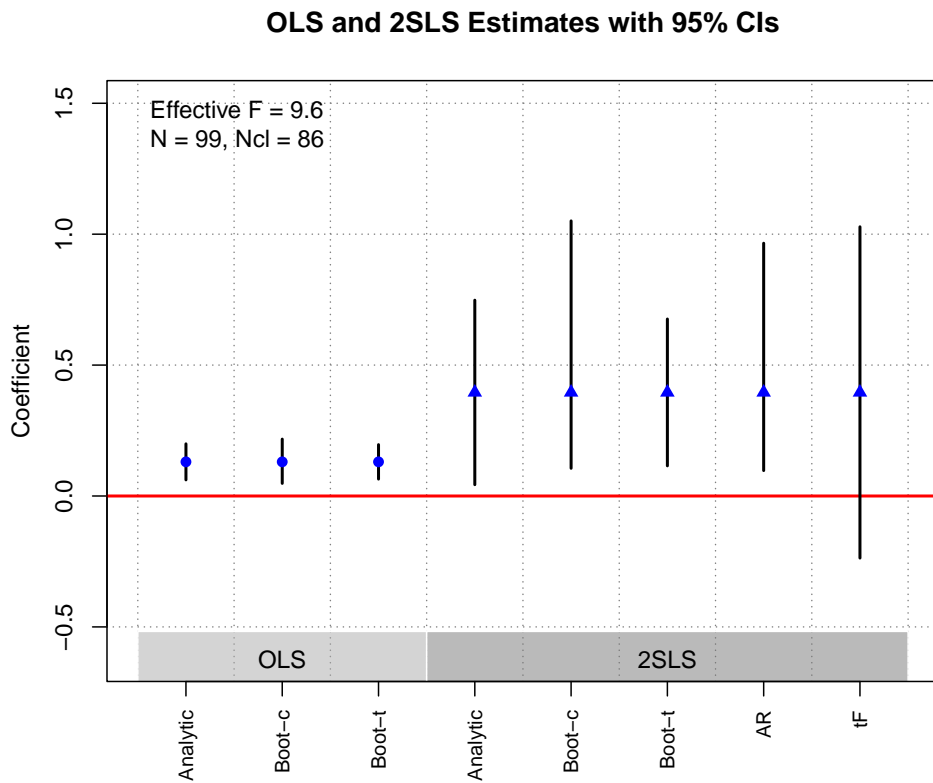
```

```

## $N
## [1] 99
##
## $N_c1
## [1] 86
##
## $df
## [1] 85
##
## $nvalues
##      gfcf_priv_gdp gov1_yrs military_first_alt
## [1,]           99      63                2
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`



Grossman et al. (2017)

Replication Summary

Unit of analysis	region * year
Treatment	government fragmentation
Instrument	the number of distinct landmasses;

Replication Summary

Outcome	length of medium and small streams; over-time variation in the number of regional governments public goods provision
Model	Table1(8)

```
df<-readRDS("./rawdata/jop_Grossman_2017.rds")
Y <- "ServicesCA"
D <- "ladminpc_15"
Z <- c("lmeanMINUSi_adminpc_16", "lmeanMINUSi_adminpc2_16",
      "herf", "herf2", "llength", "llength2")
controls <- c("lpop_1", "wdi_urban_1", "lgdppc_1", "conflict_1",
            "dpi_state_1", "p_polity2_1",
            "loilpc_1", "aid_pc_1", "al_ethnic")
cl <- "ccodecow"
FE <- "year"
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0364 0.0978 0.3721 -0.1554  0.2282  0.7098
## Boot.c   0.0364 0.1257 0.2896 -0.1900  0.3144  0.7385
## Boot.t   0.0364 0.0978 0.3721 -0.1886  0.2614  0.7165
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.4164 0.1623 2.5650  0.0982  0.7345  0.0103
## Boot.c   0.4164 0.2050 2.0315 -0.0902  0.6785  0.1402
## Boot.t   0.4164 0.1623 2.5650 -0.1088  0.9415  0.0962
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    3.8390    6.0000 511.0000 0.0009
##
## $AR$ci.print
## [1] "[0.1177, 1.3043]"
##
## $AR$ci
## [1] 0.1177 1.3043
##
## $AR$bounded
## [1] TRUE
##
```

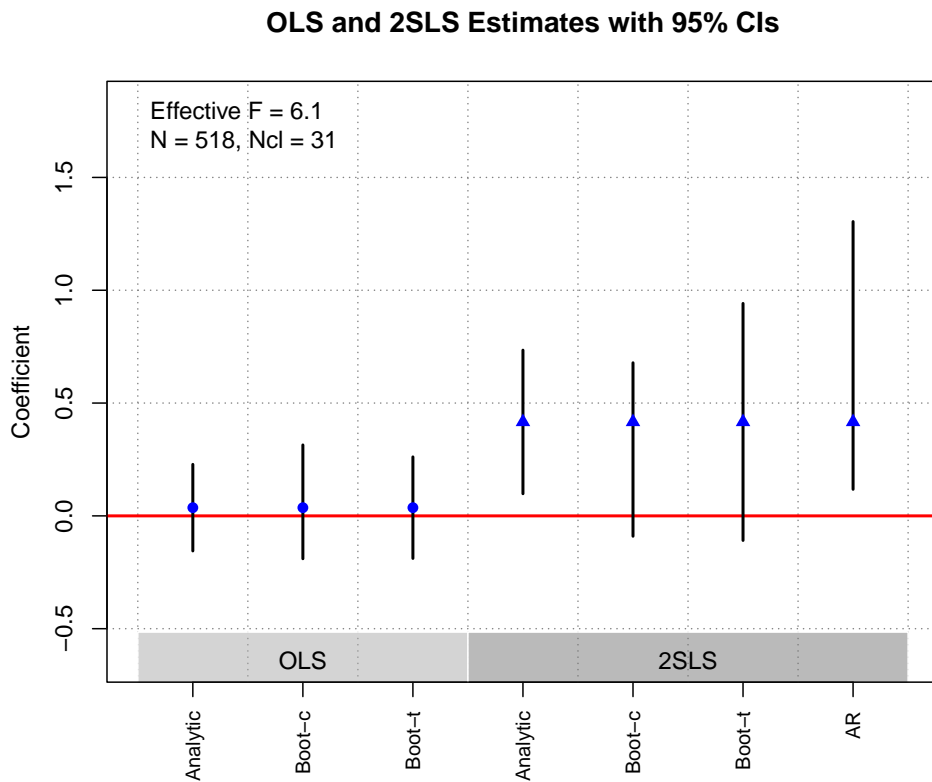
```

##
## $F_stat
## F.standard      F.robust      F.cluster F.bootstrap F.effective
##      39.9978      40.9874      11.9593      0.9983      6.1390
##
## $rho
## [1] 0.581
##
## $est_rf
##              Coef      SE p.value      SE.b  CI.b2.5% CI.b97.5%
## lmeanMINUSi_adminpc_l6  6.0801  7.3987  0.4112  12.6549  -17.7145  30.7282
## lmeanMINUSi_adminpc2_l6 -3.9097  2.3810  0.1006   3.4728  -11.0236   2.7265
## herf                    -0.0170  2.4059  0.9943  490.3141  -59.6529 1860.2308
## herf2                   -0.0545  1.7185  0.9747  252.8414  -960.7775  27.9109
## llength                  0.0669  0.0507  0.1867   1.0051  -0.9837   2.8628
## llength2                 -0.0029  0.0037  0.4309   0.0374  -0.1092   0.0363
##
##              p.value.b
## lmeanMINUSi_adminpc_l6    0.5356
## lmeanMINUSi_adminpc2_l6    0.2155
## herf                       0.7071
## herf2                      0.6590
## llength                    0.4665
## llength2                   0.5460
##
## $est_fs
##              Coef      SE p.value      SE.b  CI.b2.5% CI.b97.5%
## lmeanMINUSi_adminpc_l6  27.1296 12.2417  0.0267  22.9074  -11.2217  74.1688
## lmeanMINUSi_adminpc2_l6 -13.3452  4.9245  0.0067   7.6336  -33.1521  -2.7406
## herf                    3.5973  4.6318  0.4374  344.9919 -1245.2399  62.8496
## herf2                   -2.4844  3.1500  0.4303  178.3931  -49.9218  637.7320
## llength                  0.0536  0.0526  0.3084   0.9889  -0.8785   2.9375
## llength2                 0.0002  0.0039  0.9671   0.0367  -0.1072   0.0327
##
##              p.value.b
## lmeanMINUSi_adminpc_l6    0.1632
## lmeanMINUSi_adminpc2_l6    0.0209
## herf                       0.9770
## herf2                      0.9456
## llength                    0.4791
## llength2                   0.7887
##
## $p_iv
## [1] 6
##
## $N
## [1] 518
##
## $N_c1

```

```
## [1] 31
##
## $df
## [1] 476
##
## $nvalues
##      ServicesCA ladminpc_15 lmeanMINUSi_adminpc_16 lmeanMINUSi_adminpc2_16 herf
## [1,]          518          518                    518                    518    15
##      herf2 llength llength2
## [1,]    15        29        29
##
## attr("class")
## [1] "ivDiag"
```

`plot_coef(g)`



Healy and Malhotra (2013)

Replication Summary

Unit of analysis	individual
Treatment	the share of a respondent's siblings who are female
Instrument	whether the younger sibling is a sister
Outcome	gender-role attitude in 1973
Model	Table1(1)

```
df <- readRDS("./rawdata/jop_Healy_etal_2013.rds")
D <- "share_sis"
Y <- "womens_rights73"
Z <- "closest"
controls <- "num_sib"
cl <- "PSU"
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0451 0.0516 0.8743 -0.0561  0.1463  0.3819
## Boot.c   0.0451 0.0513 0.8800 -0.0583  0.1430  0.3800
## Boot.t   0.0451 0.0516 0.8743 -0.0307  0.1210  0.2230
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.1706 0.0844 2.0203  0.0051  0.3360  0.0434
## Boot.c   0.1706 0.0866 1.9701  0.0009  0.3462  0.0500
## Boot.t   0.1706 0.0844 2.0203  0.0470  0.2941  0.0070
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    4.1446    1.0000  277.0000  0.0427
##
## $AR$ci.print
## [1] "[0.0068, 0.3394]"
##
## $AR$ci
## [1] 0.0068 0.3394
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
##    255.3329    252.1198    244.4704    238.6458    244.4704
##
## $rho
## [1] 0.6932
##
## $tF
##           F      cF      Coef      SE      t  CI2.5%  CI97.5%  p-value
```

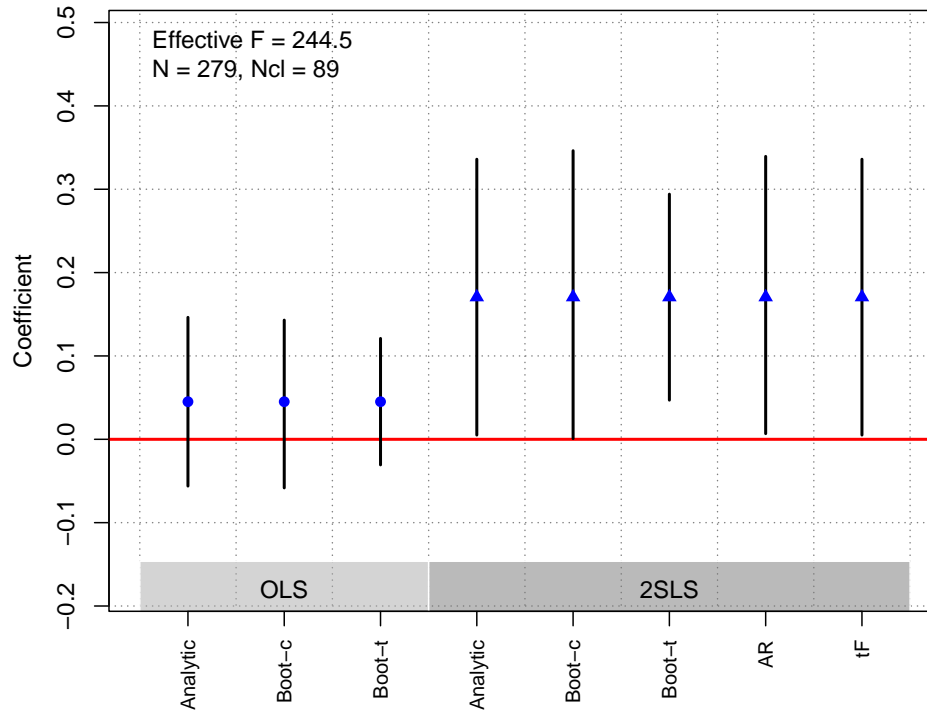
```

## 244.4704  1.9600  0.1706  0.0844  2.0203  0.0051  0.3360  0.0434
##
## $est_rf
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## closest 0.0832 0.0409  0.0421 0.0414  4e-04  0.1588  0.05
##
## $est_fs
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## closest 0.4876 0.0312  0 0.0316  0.4274  0.5508  0
##
## $p_iv
## [1] 1
##
## $N
## [1] 279
##
## $N_cl
## [1] 89
##
## $df
## [1] 276
##
## $nvalues
##      womens_rights73 share_sis closest
## [1,]                7      17      2
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



Henderson and Brooks (2016) (a)

Replication Summary

Unit of analysis	district*year
Treatment	Democratic vote margins
Instrument	rain around election day
Outcome	incumbent roll call positioning
Model	Table3(1)

```
df<- readRDS("./rawdata/jop_Henderson_etal_2016.rds")
df$fe_id_num<-df$`as.factor(fe_id_num)`
D <- "dose"
Y <- "vote"
Z <- c("rain_day", "rain_day_prev")
controls <- c("d_inc", "dist_prev", "midterm", "pres_party", "black",
             "construction", "educ", "minc", "farmer", "forborn",
             "gvtwkr", "manuf", "pop", "unempld", "urban", "retail",
             "sos", "gov", "comp_cq", "redistricted", "dose_prv", "vote_prv")
cl <- "fe_id_num" # incumbent
FE <- "fe_id_num"
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

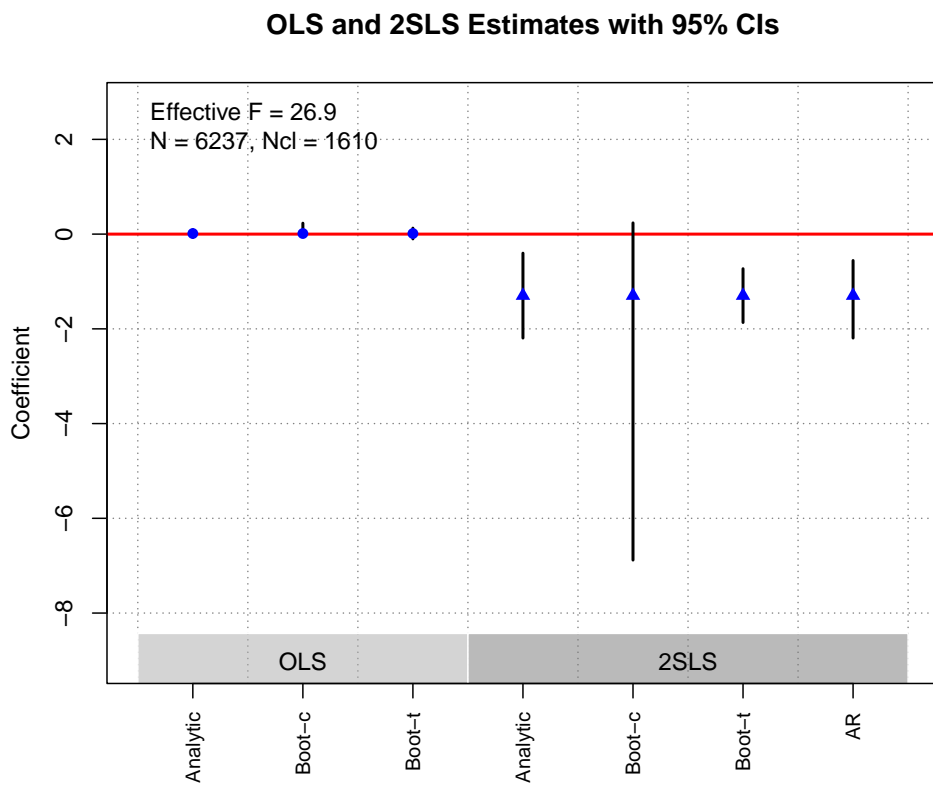
```

## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0124 0.0415 0.2996 -0.0689  0.0937  0.7645
## Boot.c   0.0124 0.0540 0.2302  0.0132  0.2308  0.0240
## Boot.t   0.0124 0.0415 0.2996 -0.1017  0.1266  0.9640
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -1.2984 0.4571 -2.8403 -2.1943 -0.4024  0.0045
## Boot.c   -1.2984 1.9367 -0.6704 -6.8812  0.2365  0.0960
## Boot.t   -1.2984 0.4571 -2.8403 -1.8661 -0.7306  0.0000
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    6.2335    2.0000 6234.0000    0.0020
##
## $AR$ci.print
## [1] "[-2.1943, -0.5578]"
##
## $AR$ci
## [1] -2.1943 -0.5578
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
##    26.4294    21.5068    22.8295    10.6450    26.9117
##
## $rho
## [1] 0.1066
##
## $est_rf
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## rain_day      0.0326 0.0100  0.0011 0.0106  0.0187  0.0595  0.000
## rain_day_prev 0.0153 0.0081  0.0585 0.0119 -0.0241  0.0224  0.996
##
## $est_fs
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## rain_day     -0.0144 0.0031      0 0.0043 -0.0192 -0.0026  0.010
## rain_day_prev -0.0187 0.0031      0 0.0045 -0.0190 -0.0015  0.016
##
## $p_iv
## [1] 2
##

```

```
## $N
## [1] 6237
##
## $N_cl
## [1] 1610
##
## $df
## [1] 1609
##
## $nvalues
##      vote dose rain_day rain_day_prev
## [1,] 6230 5138      5321          5326
##
## attr("class")
## [1] "ivDiag"
```

`plot_coef(g)`



Henderson and Brooks (2016) (b)

Replication Summary

Unit of analysis	district*year
Treatment	Democratic vote margins
Instrument	rain around election weekend

Replication Summary

Outcome incumbent roll call positioning
Model Table3(2)

```
df<- readRDS("./rawdata/jop_Henderson_etal_2016.rds")
df$fe_id_num<-df$`as.factor(fe_id_num)`
D <- "dose"
Y <- "vote"
Z <- c("rain_weekend", "rain_weekend_prev")
controls <- c("d_inc", "dist_prev", "midterm", "pres_party", "black",
             "construction", "educ", "minc", "farmer", "forborn",
             "gvtwkr", "manuf", "pop", "unempld", "urban", "retail",
             "sos", "gov", "comp_cq", "redistricted", "dose_prv", "vote_prv")
cl <- "fe_id_num" # incumbent
FE <- "fe_id_num"
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0124 0.0415 0.2996 -0.0689  0.0937  0.7645
## Boot.c   0.0124 0.0520 0.2388  0.0283  0.2285  0.0160
## Boot.t   0.0124 0.0415 0.2996 -0.1012  0.1260  0.9740
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -1.1444 0.4293 -2.6654 -1.9859 -0.3029  0.0077
## Boot.c   -1.1444 0.9276 -1.2337 -3.1487  0.5645  0.1980
## Boot.t   -1.1444 0.4293 -2.6654 -1.8740 -0.4148  0.0040
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    4.7151    2.0000 6234.0000  0.0090
##
## $AR$ci.print
## [1] "[-2.2864, -0.2685]"
##
## $AR$ci
## [1] -2.2864 -0.2685
##
## $AR$bounded
## [1] TRUE
##
##
```

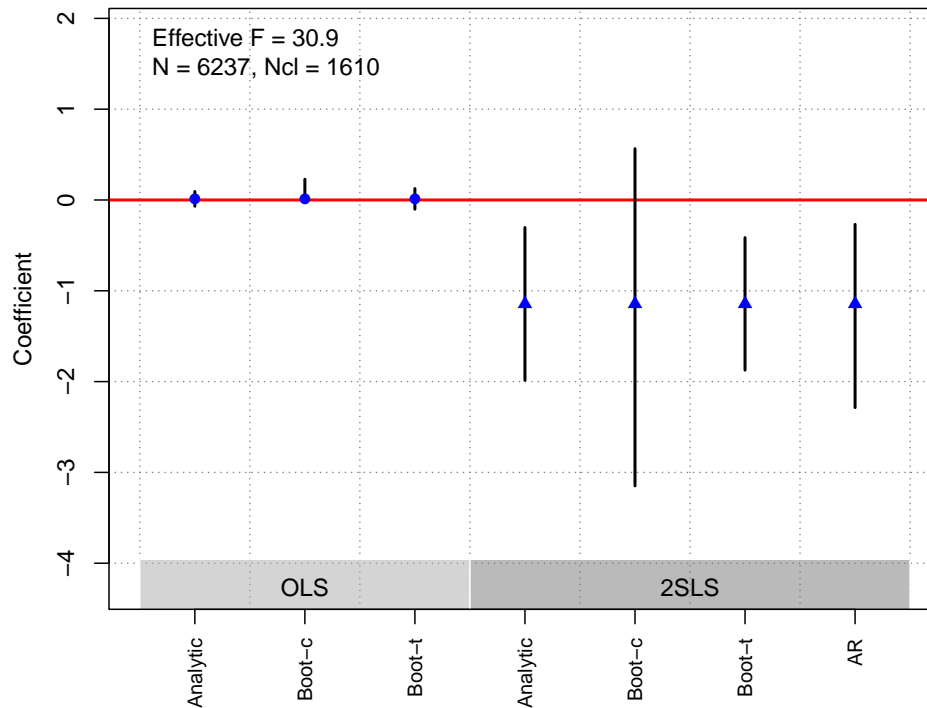
```

## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 30.3614 24.5741 26.3171 14.1875 30.9359
##
## $rho
## [1] 0.1141
##
## $est_rf
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## rain_weekend 0.0306 0.0109 0.0050 0.0116 0.0069 0.0519 0.012
## rain_weekend_prev 0.0175 0.0095 0.0665 0.0148 -0.0294 0.0268 0.954
##
## $est_fs
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## rain_weekend -0.0192 0.0034 0 0.0048 -0.0251 -0.0063 0
## rain_weekend_prev -0.0213 0.0035 0 0.0047 -0.0231 -0.0050 0
##
## $p_iv
## [1] 2
##
## $N
## [1] 6237
##
## $N_c1
## [1] 1610
##
## $df
## [1] 1609
##
## $nvalues
## vote dose rain_weekend rain_weekend_prev
## [1,] 6230 5138 5401 5407
##
## attr("class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



Johns and Pelc (2016)

Replication Summary	
Unit of analysis	WTO dispute
Treatment	the number third parties
Instrument	trade stake of the rest of the world
Outcome	becoming a third party
Model	Table2(2)

```
df<-readRDS("./rawdata/jop_Johns_etal_2016.rds")
D='third_num_excl'
Y='thirdparty'
Z='ln_ROW_before_disp'
controls=c("ln_gdpk_partner", "ln_history_third", "ln_history_C",
           "Multilateral", "trade_before_dispute", "ARTICLEXXII")
cl <- NULL
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##      Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.019 0.0017 11.3469 0.0157 0.0223 0
```

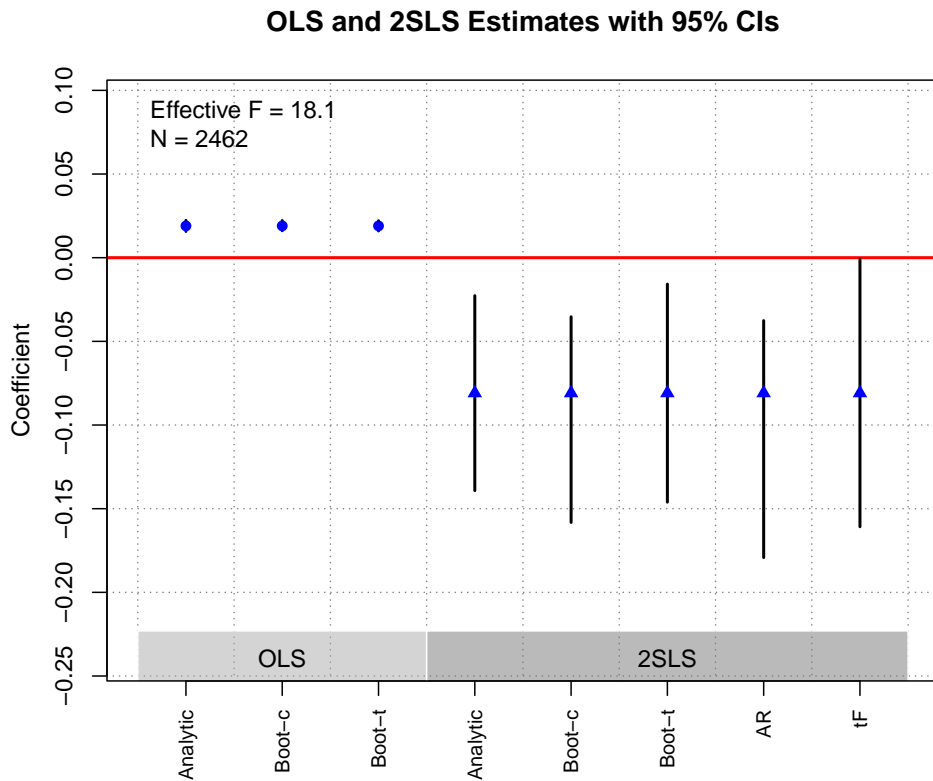
```

## Boot.c  0.019 0.0016 11.8915  0.0161  0.0222      0
## Boot.t  0.019 0.0017 11.3469  0.0160  0.0221      0
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.0809 0.0297 -2.7247 -0.1392 -0.0227 0.0064
## Boot.c   -0.0809 0.0335 -2.4134 -0.1582 -0.0353 0.0000
## Boot.t   -0.0809 0.0297 -2.7247 -0.1461 -0.0158 0.0250
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
## 19.7186  1.0000 2460.0000  0.0000
##
## $AR$ci.print
## [1] "[-0.1792, -0.0376]"
##
## $AR$ci
## [1] -0.1792 -0.0376
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
## 16.9224     18.1200      NA      18.9071     18.1200
##
## $rho
## [1] 0.0828
##
## $tF
##          F      cF      Coef      SE      t CI2.5% CI97.5% p-value
## 18.1200  2.6873 -0.0809  0.0297 -2.7247 -0.1608 -0.0011  0.0469
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## ln_ROW_before_disp -0.0137 0.0031      0 0.0031 -0.0195 -0.0076      0
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## ln_ROW_before_disp 0.1692 0.0397      0 0.0389  0.1015  0.2536      0
##
## $p_iv
## [1] 1
##
## $N

```

```
## [1] 2462
##
## $N_cl
## NULL
##
## $df
## [1] 2454
##
## $nvalues
##      thirdparty third_num_excl ln_ROW_before_disp
## [1,]          2             17             2281
##
## attr("class")
## [1] "ivDiag"
```

plot_coef(g)



Kriner and Schickler (2014)

Replication Summary

Unit of analysis	month
Treatment	committee investigations
Instrument	number of days that Congress was in session in a given month
Outcome	presidential approval

Replication Summary

Model Table1(1)

```
df<-readRDS("./rawdata/jop_Kriner_etal_2014.rds")
D <- "misconductdays"
Y <- "approval"
Z <- "alldaysinsession"
controls <- c("icst1", "positive", "negative", "vcaslast6mos",
             "iraqcaslast6mos", "honeymoon", "approvalt1", "ike","jfk",
             "lbj","rmn","ford","carter","reagan","bush","clinton","wbush")
cl <- NULL
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.0314 0.0149 -2.1103 -0.0606 -0.0022 0.0348
## Boot.c   -0.0314 0.0150 -2.0957 -0.0624 -0.0027 0.0240
## Boot.t   -0.0314 0.0149 -2.1103 -0.0612 -0.0017 0.0350
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -0.1262 0.0449 -2.8096 -0.2142 -0.0382 0.005
## Boot.c   -0.1262 0.0455 -2.7699 -0.2217 -0.0471 0.002
## Boot.t   -0.1262 0.0449 -2.8096 -0.2100 -0.0423 0.001
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    8.9171    1.0000 634.0000 0.0029
##
## $AR$ci.print
## [1] "[-0.2196, -0.0426]"
##
## $AR$ci
## [1] -0.2196 -0.0426
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
##    105.5872  121.5394          NA    124.0945    121.5394
```

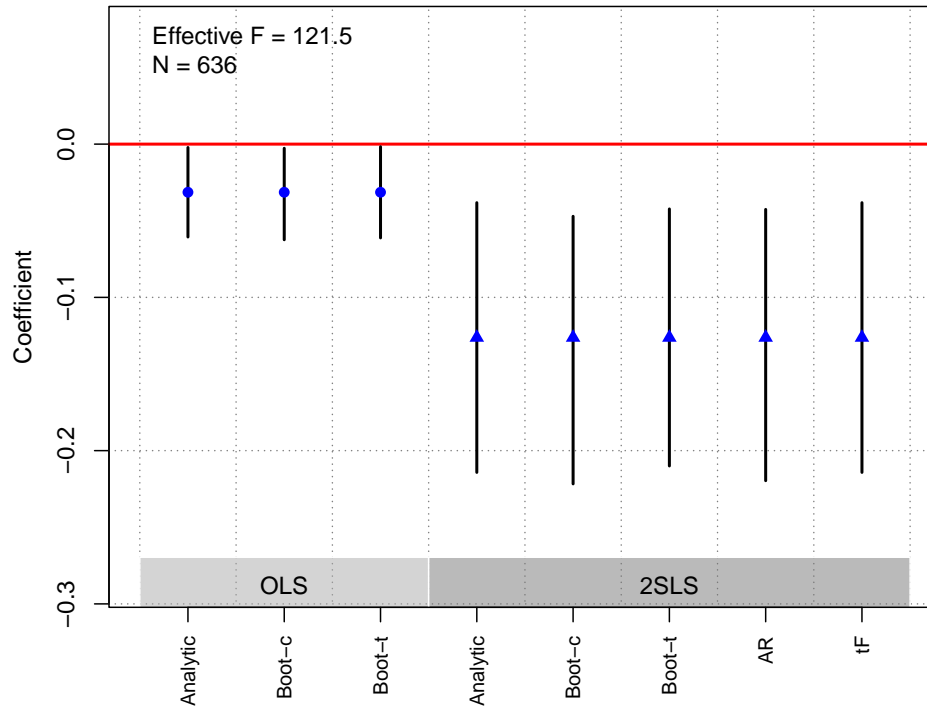
```

##
## $rho
## [1] 0.382
##
## $tF
##      F      cF      Coef      SE      t      CI2.5%  CI97.5%  p-value
## 121.5394  1.9600 -0.1262  0.0449 -2.8096 -0.2142 -0.0382  0.0050
##
## $est_rf
##              Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## alldaysinsession -0.035 0.0119  0.0032 0.0117 -0.0584 -0.0137  0.002
##
## $est_fs
##              Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## alldaysinsession 0.2777 0.0252      0 0.0249  0.226  0.3269      0
##
## $p_iv
## [1] 1
##
## $N
## [1] 636
##
## $N_cl
## NULL
##
## $df
## [1] 618
##
## $nvalues
##      approval misconductdays alldaysinsession
## [1,]      185              52              49
##
## attr(,"class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



Lei and Zhou (2022)

Replication Summary

Unit of analysis	city*year
Treatment	subway approval
Instrument	whether the city has more than 3 million residents*
Outcome	population size
Model	mayor promotion
	Table3(A)

```
df<-readRDS("./rawdata/jop_Lei_2022.rds")
Y <- 'Mayor_promotion3y'
D <- 'Mayor_plan'
Z <- 'iv1'
controls<-c( 'Per_pop_2', 'iv1_int')
cl<-"City_Code"
FE<-c("provinceyear", "City_Code")
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##      Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.276 0.1196 2.3077 0.0416 0.5104 0.0210
```

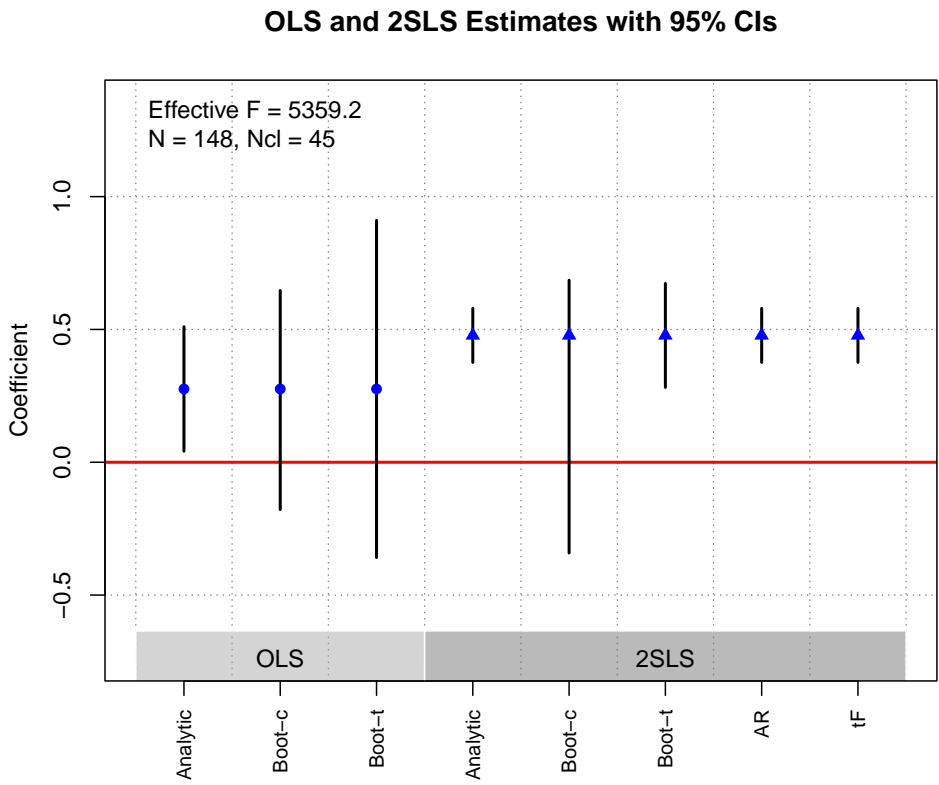
```

## Boot.c  0.276 0.2244 1.2299 -0.1781  0.6467  0.1372
## Boot.t  0.276 0.1196 2.3077 -0.3587  0.9106  0.2955
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.4776 0.0519 9.2026  0.3759  0.5793  0.0000
## Boot.c   0.4776 0.2701 1.7681 -0.3416  0.6852  0.1478
## Boot.t   0.4776 0.0519 9.2026  0.2820  0.6731  0.0026
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
## 83.1817  1.0000 146.0000  0.0000
##
## $AR$ci.print
## [1] "[0.3759, 0.5793]"
##
## $AR$ci
## [1] 0.3759 0.5793
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
## 53.4747  2276.8055  5359.1714  181.9936  5359.1714
##
## $rho
## [1] 0.7604
##
## $tF
##          F      cF      Coef      SE      t      CI2.5%  CI97.5%  p-value
## 5359.1714  1.9600  0.4776  0.0519  9.2026  0.3759  0.5793  0.0000
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## iv1 0.4833 0.0534      0 0.2797 -0.357  0.6955  0.1478
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## iv1 1.0119 0.0138      0 0.075  0.9858  1.2165      0
##
## $p_iv
## [1] 1
##
## $N

```

```
## [1] 148
##
## $N_cl
## [1] 45
##
## $df
## [1] 39
##
## $nvalues
##      Mayor_promotion3y Mayor_plan iv1
## [1,]                2          2   2
##
## attr("class")
## [1] "ivDiag"
```

plot_coef(g)



Lerman et al. (2017)

Replication Summary

Unit of analysis	individual
Treatment	public versus only private health insurance
Instrument	born 1946 or 1947
Outcome	support ACA

Replication Summary

Model Table1(1)

```
df<-readRDS("./rawdata/jop_Lerman_2017.rds")
Y <- 'suppafford'
D <- 'privpubins3r'
Z <- 'byr4647'
controls<-c( 'rep', 'ind', 'con', 'mod',
             'ideostrength', 'hcsocial', 'fininsur',
             'healthcaresupport', 'child18', 'male',
             'married', 'labor', 'mobility', 'homeowner',
             'religimp', 'employed', 'votereg', 'vote08',
             'black', 'hispanic2', 'military', 'educ',
             'fincome', 'newsint', 'publicemp', 'bornagain')
cl<-NULL
FE<-NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0093 0.0109 0.8542 -0.0121  0.0307  0.393
## Boot.c   0.0093 0.0109 0.8578 -0.0117  0.0298  0.414
## Boot.t   0.0093 0.0109 0.8542 -0.0116  0.0302  0.402
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0459 0.0229 2.0095  0.0011  0.0908  0.0445
## Boot.c   0.0459 0.0230 1.9971  0.0033  0.0909  0.0320
## Boot.t   0.0459 0.0229 2.0095  0.0027  0.0892  0.0360
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    4.0770    1.0000 4387.0000  0.0435
##
## $AR$ci.print
## [1] "[0.0016, 0.0908]"
##
## $AR$ci
## [1] 0.0016 0.0908
##
## $AR$bounded
## [1] TRUE
##
```

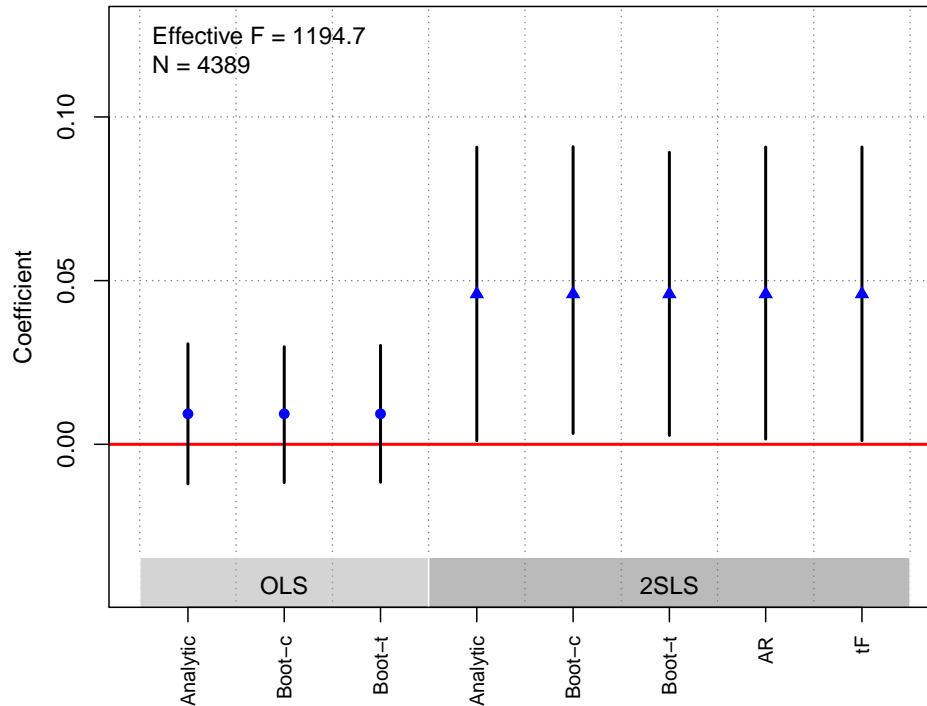
```

##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 1272.162 1194.659 NA 1143.951 1194.659
##
## $rho
## [1] 0.4752
##
## $tF
## F cF Coef SE t CI2.5% CI97.5% p-value
## 1194.6594 1.9600 0.0459 0.0229 2.0095 0.0011 0.0908 0.0445
##
## $est_rf
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## byr4647 0.0202 0.01 0.0441 0.0101 0.0015 0.0396 0.032
##
## $est_fs
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## byr4647 0.4401 0.0127 0 0.013 0.4124 0.4645 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 4389
##
## $N_cl
## NULL
##
## $df
## [1] 4361
##
## $nvalues
## suppafford privpubins3r byr4647
## [1,] 2 2 2
##
## attr(,"class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



Lorentzen et al. (2014)

Replication Summary

Unit of analysis	city
Treatment	large firm dominance in 2007
Instrument	same variable measured in 1999
Outcome	pollution information transparency index
Model	Table1(2)

```
df<-readRDS("./rawdata/jop_Lorentzen_2014.rds")
D <- "lfd2007"
Y <- "pitiave3"
Z <- "lfd99"
controls <- c("lbudgetrev", "lexpratio", "tertratio", "sat_air_pca")
cl <- NULL
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -2.4789 1.0508 -2.3590 -4.5385 -0.4193 0.0183
## Boot.c   -2.4789 1.0523 -2.3557 -4.3565 -0.3374 0.0140
```

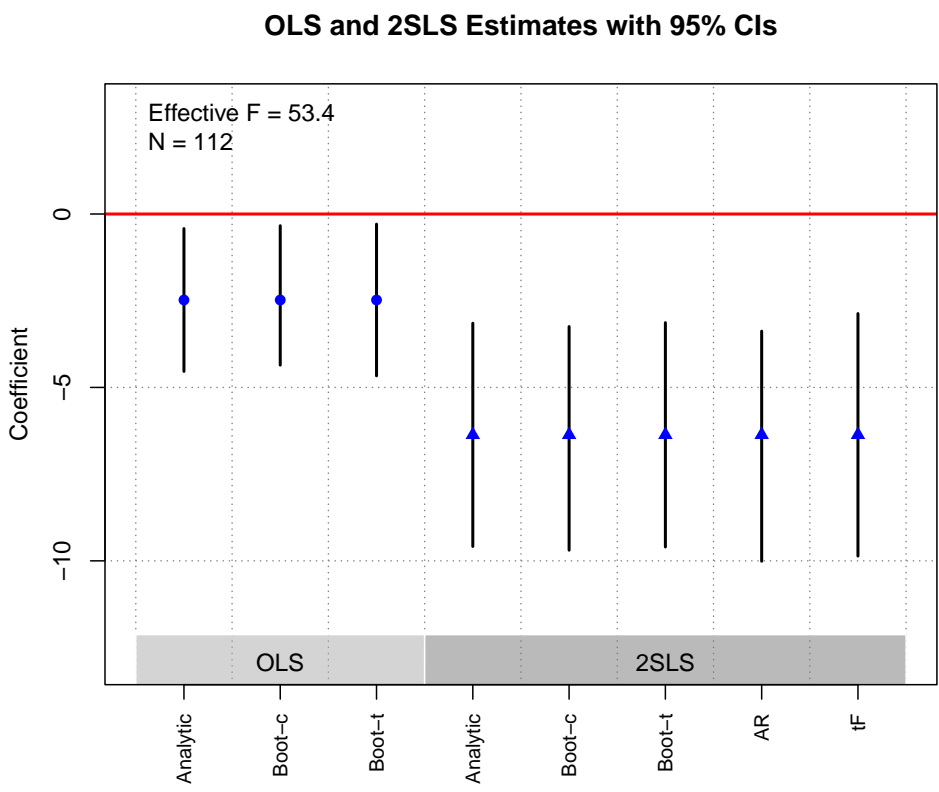
```

## Boot.t   -2.4789  1.0508 -2.3590 -4.6666  -0.2911  0.0280
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic -6.3664  1.6421 -3.8769 -9.5850  -3.1478  1e-04
## Boot.c   -6.3664  1.6629 -3.8285 -9.6929  -3.2453  2e-03
## Boot.t   -6.3664  1.6421 -3.8769 -9.6001  -3.1328  1e-03
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##  17.3155  1.0000 110.0000  0.0001
##
## $AR$ci.print
## [1] "[-10.0120, -3.3777]"
##
## $AR$ci
## [1] -10.0120  -3.3777
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
##   53.6182    53.4100      NA      57.4805    53.4100
##
## $rho
## [1] 0.5796
##
## $tF
##           F      cF      Coef      SE      t CI2.5% CI97.5% p-value
##  53.4100  2.1292 -6.3664  1.6421 -3.8769 -9.8628 -2.8700  0.0004
##
## $est_rf
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## lfd99 -3.4227  0.8379      0 0.8463  -4.9716  -1.7773      0.002
##
## $est_fs
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## lfd99  0.5376  0.0736      0 0.0709   0.4093   0.6829      0
##
## $p_iv
## [1] 1
##
## $N
## [1] 112

```

```
##
## $N_cl
## NULL
##
## $df
## [1] 106
##
## $nvalues
##      pitiave3 lfd2007 lfd99
## [1,]      108      112  112
##
## attr("class")
## [1] "ivDiag"
```

plot_coef(g)



Pianzola et al. (2019)

Replication Summary

Unit of analysis	individual
Treatment	smartvote use
Instrument	random assignment of the e-mail treatment
Outcome	vote intentions
Model	Table4(3)

```
df <- readRDS("./rawdata/jop_Pianzola_etal_2019.rds")
D <- "smartvote"
Y <- "diff_top_ptv"
Z <- "email"
controls <- NULL
cl <- NULL
FE <- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0805 0.0684 1.1767 -0.0536  0.2146  0.2393
## Boot.c   0.0805 0.0676 1.1913 -0.0510  0.2228  0.2340
## Boot.t   0.0805 0.0684 1.1767 -0.0534  0.2144  0.2180
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.755 0.3788 1.9934  0.0126  1.4974  0.0462
## Boot.c   0.755 0.3811 1.9810  0.0701  1.5729  0.0320
## Boot.t   0.755 0.3788 1.9934  0.0486  1.4615  0.0380
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    4.2767    1.0000 1773.0000  0.0388
##
## $AR$ci.print
## [1] "[0.0429, 1.5883]"
##
## $AR$ci
## [1] 0.0429 1.5883
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
##    46.7293    46.7612         NA    48.7842    46.7612
##
## $rho
## [1] 0.1602
##
## $tF
##           F      cF      Coef      SE      t CI2.5% CI97.5% p-value
```

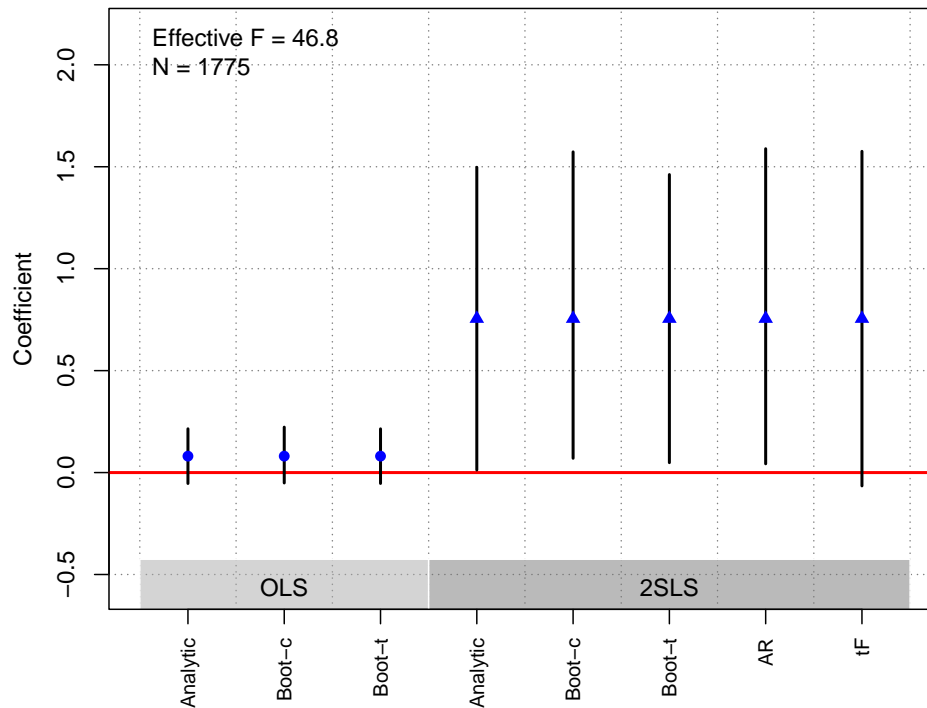
```

## 46.7612  2.1662  0.7550  0.3788  1.9934 -0.0654  1.5755  0.0713
##
## $est_rf
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## email 0.1032 0.0499  0.0386 0.0484  0.0091  0.1989  0.032
##
## $est_fs
##      Coef  SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## email 0.1367 0.02      0 0.0196  0.0992  0.1762      0
##
## $p_iv
## [1] 1
##
## $N
## [1] 1775
##
## $N_c1
## NULL
##
## $df
## [1] 1773
##
## $nvalues
##      diff_top_ptv smartvote email
## [1,]           18           2    2
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



Schleiter and Tavits (2016)

Replication Summary

Unit of analysis	election
Treatment	opportunistic election calling
Instrument	prime Minister dissolution power
Outcome	vote share of Prime Minister's party
Model	Table3(b4)

```
df<- readRDS("./rawdata/jop_Schleiter_etal_2016.rds")
D <- "term2"
Y <- "pm_voteshare_next"
Z <- "disspm"
controls <- c("pm_voteshare", "gdp_chg1yr", "cpi1yr", "dumcpi1yr")
cl <- "countryn"
FE <- "decade"
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 3.0828 1.0369 2.9730 1.0504  5.1152  0.0029
## Boot.c   3.0828 1.1561 2.6665 1.4578  5.9719  0.0000
```

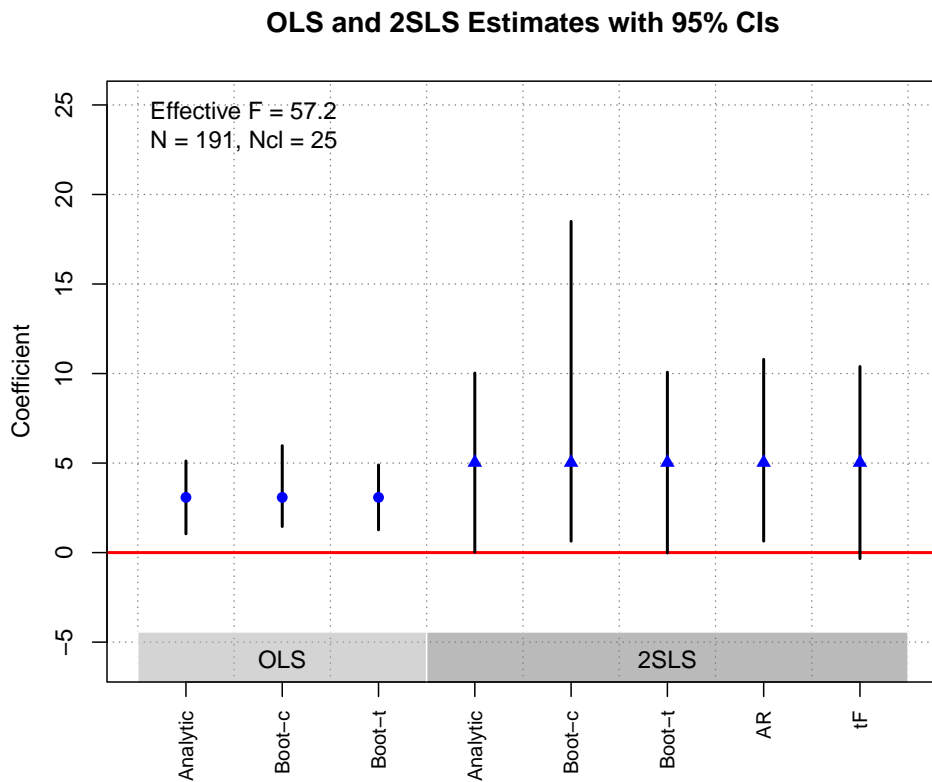
```

## Boot.t   3.0828 1.0369 2.9730  1.2722  4.8933  0.0060
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 5.0282  2.5494 1.9723  0.0314  10.0250  0.0486
## Boot.c   5.0282 67.7447 0.0742  0.6404  18.4948  0.0340
## Boot.t   5.0282  2.5494 1.9723 -0.0190  10.0754  0.0510
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##   5.1692  1.0000 189.0000  0.0241
##
## $AR$ci.print
## [1] "[0.6433, 10.7899]"
##
## $AR$ci
## [1] 0.6433 10.7899
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
##   107.0322    75.6881    57.1949    23.5901    57.1949
##
## $rho
## [1] 0.6117
##
## $tF
##           F      cF      Coef      SE      t CI2.5% CI97.5% p-value
##   57.1949  2.1037  5.0282  2.5494  1.9723 -0.3350 10.3914  0.0661
##
## $est_rf
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## disspm 0.3124 0.1412 0.0269 0.1749  0.0746  0.7642  0.012
##
## $est_fs
##           Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## disspm 0.0621 0.0082  0 0.0128  0.0209  0.0746  0.022
##
## $p_iv
## [1] 1
##
## $N
## [1] 191

```

```
##
## $N_cl
## [1] 25
##
## $df
## [1] 179
##
## $nvalues
##      pm_voteshare_next term2 disspm
## [1,]           157      2      6
##
## attr("class")
## [1] "ivDiag"
```

`plot_coef(g)`



Schubiger (2021)

Replication Summary

Unit of analysis	community
Treatment	exposure to state violence
Instrument	location of a community inside or outside the emergency zone
Outcome	counterinsurgent mobilization

```

df <-readRDS("./rawdata/jop_Schubiger_2021.rds")
D <- "violence_est_period2"
Y<-"autodefensa"
Z <- "emzone"
controls <- "distance"
cl<- NULL
FE<- NULL
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))

```

```

## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0702 0.014 5.0069 0.0427 0.0977 0
## Boot.c   0.0702 0.014 5.0115 0.0454 0.0989 0
## Boot.t   0.0702 0.014 5.0069 0.0419 0.0986 0
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.2736 0.0764 3.5814 0.1239 0.4234 3e-04
## Boot.c   0.2736 0.0755 3.6254 0.1390 0.4280 0e+00
## Boot.t   0.2736 0.0764 3.5814 0.1319 0.4154 1e-03
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
## 12.7351 1.0000 7293.0000 0.0004
##
## $AR$ci.print
## [1] "[0.1300, 0.4463]"
##
## $AR$ci
## [1] 0.1300 0.4463
##
## $AR$bounded
## [1] TRUE
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 39.9899 38.5348 NA 40.7010 38.5348
##
## $rho
## [1] 0.0739
##
## $tF
##           F      cF      Coef      SE      t CI2.5% CI97.5% p-value

```

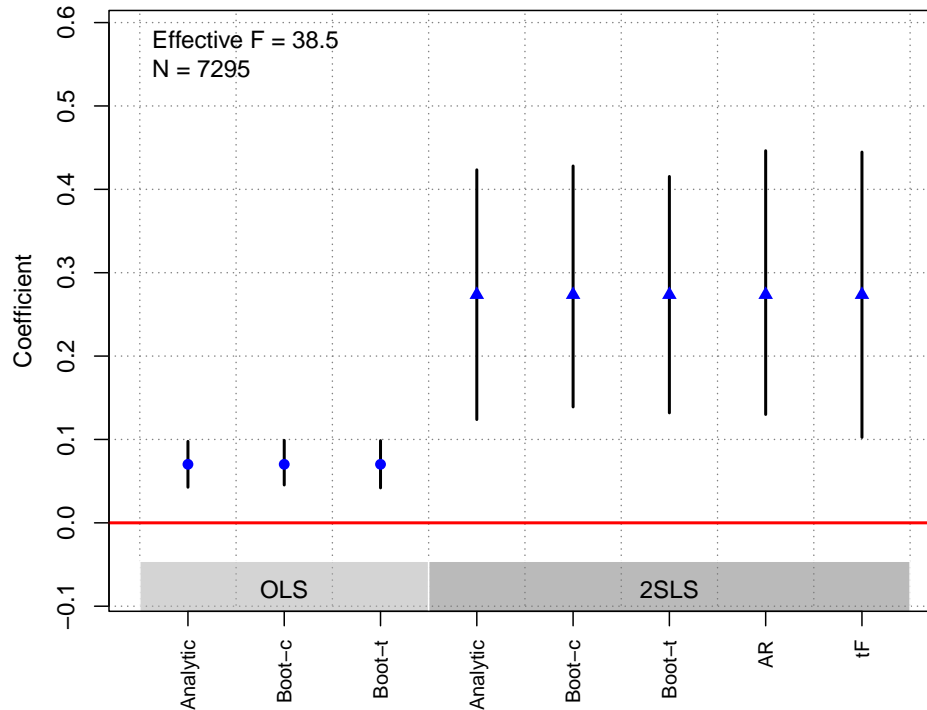
```

## 38.5348  2.2392  0.2736  0.0764  3.5814  0.1025  0.4447  0.0017
##
## $est_rf
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## emzone 0.0172 0.0048 4e-04 0.0047  0.0084  0.027  0
##
## $est_fs
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## emzone 0.0629 0.0101  0 0.0099  0.044  0.0834  0
##
## $p_iv
## [1] 1
##
## $N
## [1] 7295
##
## $N_c1
## NULL
##
## $df
## [1] 7292
##
## $nvalues
##      autodefensa violence_est_period2 emzone
## [1,]          2                2      2
##
## attr("class")
## [1] "ivDiag"

```

`plot_coef(g)`

OLS and 2SLS Estimates with 95% CIs



Stewart and Liou (2017)

Replication Summary

Unit of analysis	insurgency*year
Treatment	foreign territory
Instrument	log total border length and the total number of that state's neighbors
Outcome	civilian casualties
Model	Table3(1)

```
df <- readRDS("./rawdata/jop_Stewart_2017.rds")
D <- "extrrdum_low"
Y <- "oneside_best_log"
Z <- "total_border_ln"
controls <- c("bd_log", "terrdu", "strengthcent_ord", "rebstrength_ord",
             'nonmilsupport', 'rebestsize', 'l1popdensity',
             'l1gdppc_log', 'l1gdppc_change')
cl <- NULL
FE <- c("year", "countrynum")
weights <- NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl, weights=weights, cores = cores))
```

```
## $est_ols
##          Coef      SE      t CI 2.5% CI 97.5% p.value
```

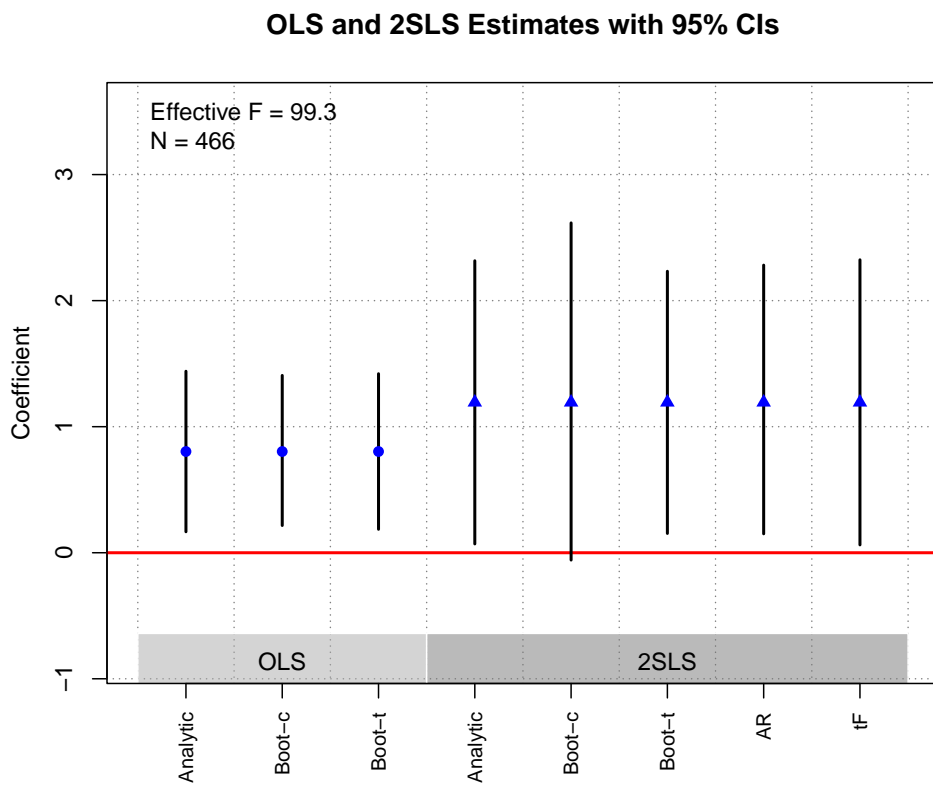
```

## Analytic 0.803 0.3249 2.4716 0.1662 1.4398 0.0135
## Boot.c 0.803 0.3153 2.5465 0.2160 1.4065 0.0140
## Boot.t 0.803 0.3249 2.4716 0.1859 1.4201 0.0120
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 1.1929 0.5730 2.0817 0.0698 2.3161 0.0374
## Boot.c 1.1929 1.4502 0.8226 -0.0582 2.6166 0.0640
## Boot.t 1.1929 0.5730 2.0817 0.1530 2.2328 0.0270
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
## 5.0089 1.0000 464.0000 0.0257
##
## $AR$ci.print
## [1] "[0.1500, 2.2817]"
##
## $AR$ci
## [1] 0.1500 2.2817
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
## 33.9859 99.3150 NA 55.3927 99.3150
##
## $rho
## [1] 0.2786
##
## $tF
##          F      cF      Coef      SE      t CI2.5% CI97.5% p-value
## 99.3150 1.9734 1.1929 0.5730 2.0817 0.0621 2.3238 0.0387
##
## $est_rf
##          Coef      SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## total_border_ln -7.0905 3.3952 0.0368 6.1515 -15.0568 0.3445 0.064
##
## $est_fs
##          Coef      SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## total_border_ln -5.9438 0.5964 0 0.7986 -7.2768 -4.5464 0
##
## $p_iv
## [1] 1
##

```

```
## $N
## [1] 466
##
## $N_c1
## NULL
##
## $df
## [1] 404
##
## $nvalues
##      onside_best_log  exterrdum_low  total_border_ln
## [1,]                113              2                45
##
## attr("class")
## [1] "ivDiag"
```

`plot_coef(g)`



Urpelainen and Zhang (2022)

Replication Summary

Unit of analysis
Treatment

district*year
wind turbine capacity

Replication Summary

Instrument	time trend multiplied by the wind resource of the electoral district
Outcome	Democratic vote
Model	Table3(B1)

```
df <-readRDS("./rawdata/jop_urpelainen_2022.rds")
D <- "cum_capacity_turbine"
Y<-"demvotesmajorpercent"
Z <- "inter"
controls <-NULL
cl<- "district_fixed"
FE<- c("stateyear_fixed","district_fixed")
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0063 0.0028 2.2395 8e-04 0.0118 0.0251
## Boot.c   0.0063 0.0035 1.8262 1e-04 0.0138 0.0460
## Boot.t   0.0063 0.0028 2.2395 3e-04 0.0123 0.0380
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0296 0.0109 2.7312 0.0084 0.0509 0.0063
## Boot.c   0.0296 0.0151 1.9612 0.0111 0.0674 0.0020
## Boot.t   0.0296 0.0109 2.7312 0.0123 0.0469 0.0030
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    9.5546    1.0000 1142.0000 0.0020
##
## $AR$ci.print
## [1] "[0.0112, 0.0618]"
##
## $AR$ci
## [1] 0.0112 0.0618
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard F.robust F.cluster F.bootstrap F.effective
```

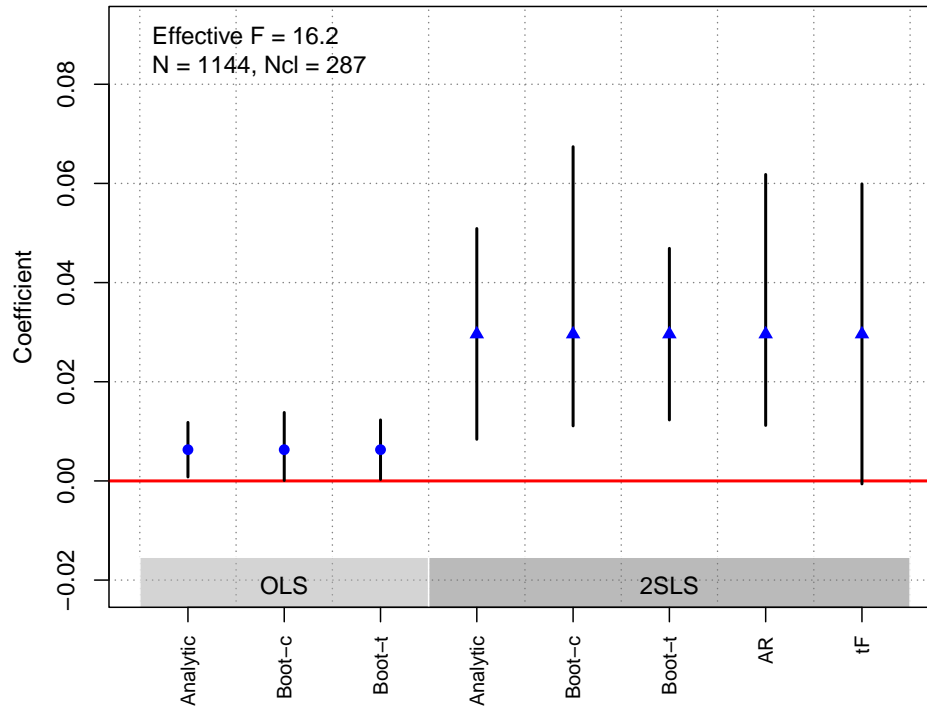
```

##      93.4366      27.8543      16.1654      15.2135      16.1654
##
## $rho
## [1] 0.3269
##
## $tF
##      F      cF      Coef      SE      t      CI2.5% CI97.5% p-value
## 16.1654 2.7897 0.0296 0.0109 2.7312 -0.0006 0.0599 0.0550
##
## $est_rf
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## inter 0.9095 0.2942 0.002 0.3123 0.2783 1.4911 0.002
##
## $est_fs
##      Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## inter 30.6883 7.6327 1e-04 7.8679 13.5585 43.7354 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 1144
##
## $N_c1
## [1] 287
##
## $df
## [1] 286
##
## $nvalues
##      demvotesmajorpercent cum_capacity_turbine inter
## [1,]                965                141 777
##
## attr("class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



Webster et al. (2022)

Replication Summary

Unit of analysis	individual
Treatment	percentage of angry words that a respondent wrote in his or her emotional recall prompt
Instrument	treatment assignment indicator
Outcome	social polarization: do favors
Model	Table2(1)

```
df <- readRDS("./rawdata/jop_Webster_2022.rds")
D <- "anger"
Y <- "fourpack_1_01"
Z <- "treated"
controls <- "democrat"
cl <- NULL
FE <- NULL
weights <- NULL
(g <- ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl, weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0024 0.0018 1.3413 -0.0011  0.0058 0.1798
```

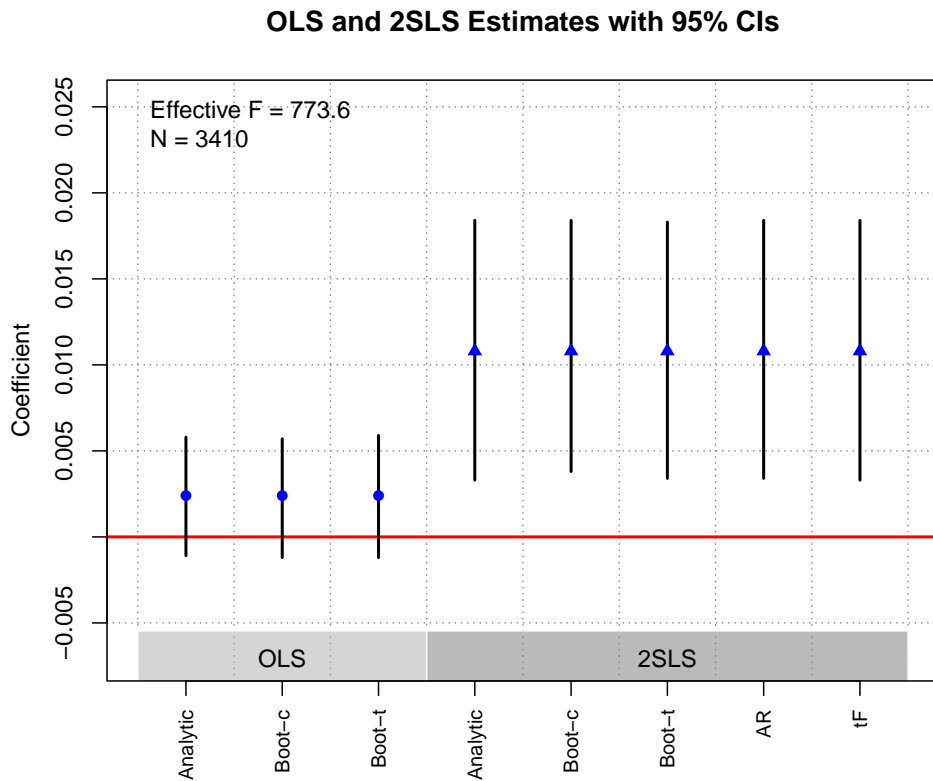
```

## Boot.c  0.0024 0.0018 1.3282 -0.0012  0.0057  0.2100
## Boot.t  0.0024 0.0018 1.3413 -0.0012  0.0059  0.1920
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0108 0.0039 2.8123  0.0033  0.0184  0.0049
## Boot.c   0.0108 0.0038 2.8797  0.0038  0.0184  0.0000
## Boot.t   0.0108 0.0039 2.8123  0.0034  0.0183  0.0000
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
##   7.9872    1.0000 3408.0000  0.0047
##
## $AR$ci.print
## [1] "[0.0034, 0.0184]"
##
## $AR$ci
## [1] 0.0034 0.0184
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster  F.bootstrap  F.effective
##   801.9232   773.5894         NA    804.5041    773.5894
##
## $rho
## [1] 0.4365
##
## $tF
##          F      cF      Coef      SE      t  CI2.5%  CI97.5%  p-value
##  773.5894  1.9600  0.0108  0.0039  2.8123  0.0033  0.0184  0.0049
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## treated 0.031 0.011  0.0047 0.0107  0.0107  0.0529  0
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## treated 2.8585 0.1028  0 0.1008  2.677  3.0659  0
##
## $p_iv
## [1] 1
##
## $N

```

```
## [1] 3410
##
## $N_c1
## NULL
##
## $df
## [1] 3407
##
## $nvalues
##      fourpack_1_01 anger treated
## [1,]           5    252         2
##
## attr("class")
## [1] "ivDiag"
```

plot_coef(g)



West (2017)

Replication Summary

Unit of analysis	individual
Treatment	Obama win
Instrument	IEM (prediction market) price
Outcome	political efficacy

Replication Summary

Model Table1(4)

```
df<- readRDS("./rawdata/jop_West_2017.rds")
D <- "obama"
Y <- "newindex"
Z <- "avgprice"
controls <- c("partyd1", "partyd2", "partyd3",
             "partyd4", "partyd5", "wa01_a", "wa02_a",
             "wa03_a", "wa04_a", "wa05_a", "wfc02_a",
             "ra01_b", "rd01", "wd02_b", "rkey",
             "wave_1", "dt_w12", "dt_w12_2")
cl <- NULL
FE <- c("state","religion")
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
           cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.0358 0.0112 3.2084 0.0139 0.0577 0.0013
## Boot.c   0.0358 0.0117 3.0700 0.0130 0.0573 0.0040
## Boot.t   0.0358 0.0112 3.2084 0.0139 0.0577 0.0040
##
## $est_2sls
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.2073 0.0873 2.3758 0.0363 0.3784 0.0175
## Boot.c   0.2073 0.0924 2.2446 0.0513 0.4116 0.0140
## Boot.t   0.2073 0.0873 2.3758 0.0435 0.3711 0.0110
##
## $AR
## $AR$Fstat
##           F      df1      df2      p
##    6.5244    1.0000 2281.0000 0.0107
##
## $AR$ci.print
## [1] "[0.0485, 0.4046]"
##
## $AR$ci
## [1] 0.0485 0.4046
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
```

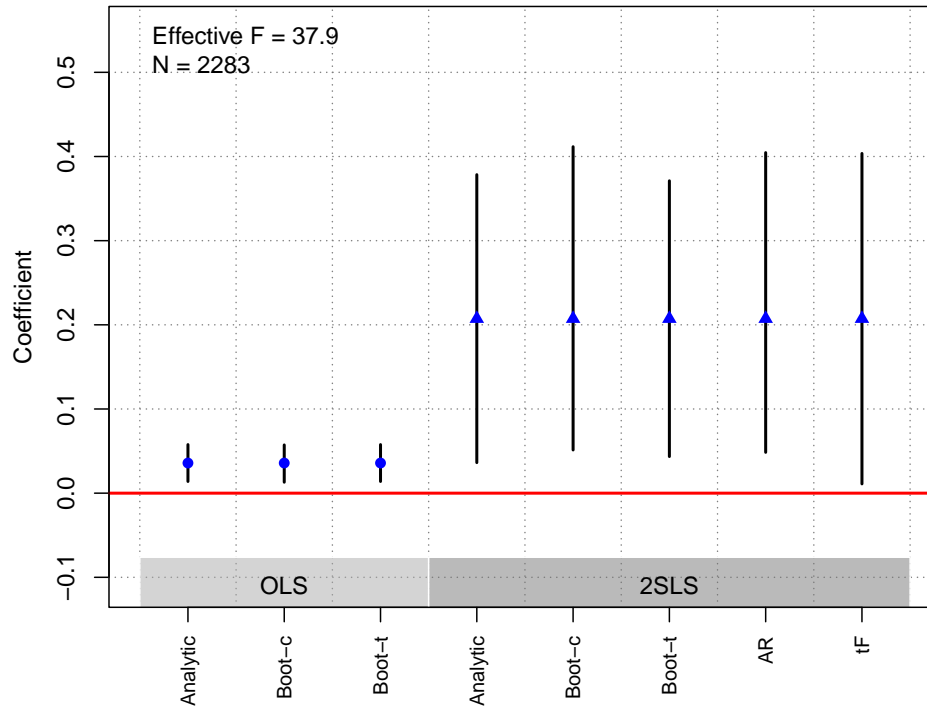
```

## F.standard F.robust F.cluster F.bootstrap F.effective
## 41.7917 37.8652 NA 39.2950 37.8652
##
## $rho
## [1] 0.1362
##
## $tF
## F cF Coef SE t CI2.5% CI97.5% p-value
## 37.8652 2.2493 0.2073 0.0873 2.3758 0.0110 0.4036 0.0384
##
## $est_rf
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## avgprice 0.1407 0.0559 0.0119 0.0559 0.0337 0.2397 0.014
##
## $est_fs
## Coef SE p.value SE.b CI.b2.5% CI.b97.5% p.value.b
## avgprice 0.6784 0.1103 0 0.1082 0.4551 0.8873 0
##
## $p_iv
## [1] 1
##
## $N
## [1] 2283
##
## $N_cl
## NULL
##
## $df
## [1] 2211
##
## $nvalues
## newindex obama avgprice
## [1,] 122 2 141
##
## attr("class")
## [1] "ivDiag"

```

```
plot_coef(g)
```

OLS and 2SLS Estimates with 95% CIs



Ziaja (2020)

Replication Summary

Unit of analysis	country*year
Treatment	number of democracy donors
Instrument	constructed instrument
Outcome	democracy scores
Model	Table1(B2)

```
df <-readRDS("./rawdata/jop_Ziaja_2020.rds")
D <- "l.CMgnh"
Y <- "v2x.polyarchy.n"
Z <- "l.ZwvCMgwh94"
controls <-c("l.pop.log.r", "l.gdpcap.log.r", "l.war25")
cl<- "cnamef"
FE<- c("cnamef", "periodf")
weights<-NULL
(g<-ivDiag(data=df, Y=Y, D=D, Z=Z, controls=controls, FE =FE,
  cl =cl,weights=weights, cores = cores))
```

```
## $est_ols
##           Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.8746 0.1931 4.5298 0.4962 1.2531      0
## Boot.c   0.8746 0.1941 4.5058 0.4727 1.2175      0
```

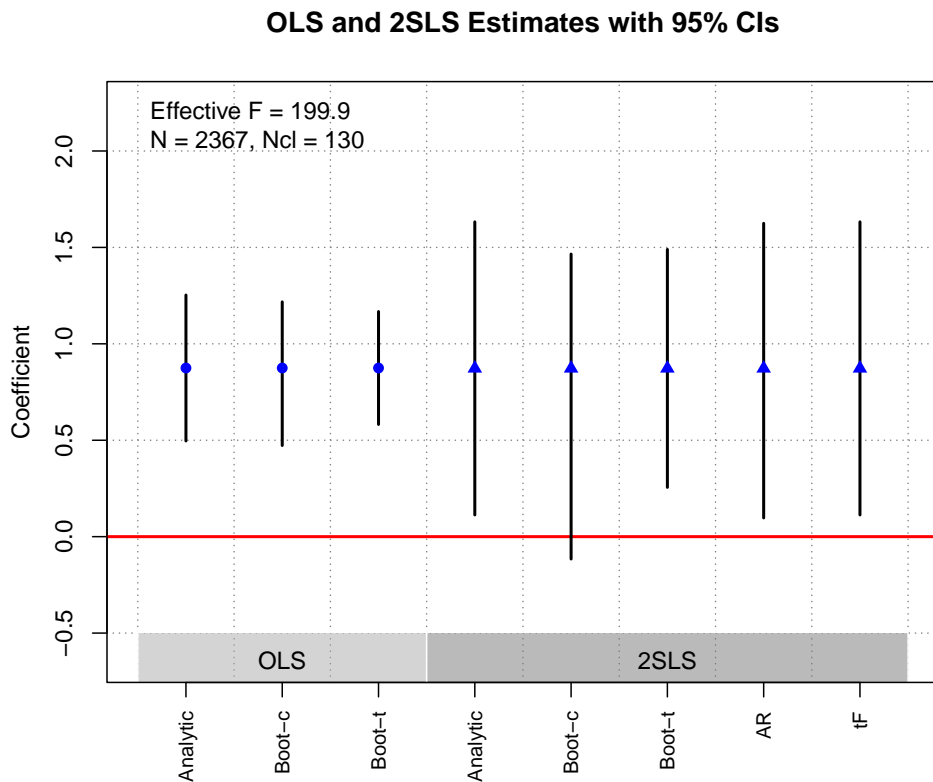
```

## Boot.t  0.8746 0.1931 4.5298  0.5819  1.1673      0
##
## $est_2sls
##          Coef      SE      t CI 2.5% CI 97.5% p.value
## Analytic 0.8726 0.3877 2.2505  0.1126  1.6325  0.0244
## Boot.c   0.8726 0.4022 2.1698 -0.1160  1.4654  0.1000
## Boot.t   0.8726 0.3877 2.2505  0.2556  1.4896  0.0030
##
## $AR
## $AR$Fstat
##          F      df1      df2      p
##   4.8018   1.0000 2365.0000  0.0285
##
## $AR$ci.print
## [1] "[0.0971, 1.6248]"
##
## $AR$ci
## [1] 0.0971 1.6248
##
## $AR$bounded
## [1] TRUE
##
##
## $F_stat
## F.standard  F.robust  F.cluster F.bootstrap F.effective
## 1158.1467   775.0850   199.9166   199.1035   199.9166
##
## $rho
## [1] 0.586
##
## $tF
##          F      cF      Coef      SE      t  CI2.5% CI97.5% p-value
## 199.9166  1.9600  0.8726  0.3877  2.2505  0.1126  1.6325  0.0244
##
## $est_rf
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## 1.ZwvCMgwh94 0.0599 0.0273 0.0285 0.0292 -0.0075  0.1064  0.1
##
## $est_fs
##          Coef      SE p.value  SE.b CI.b2.5% CI.b97.5% p.value.b
## 1.ZwvCMgwh94 0.0686 0.0049  0 0.0049  0.0614  0.0807  0
##
## $p_iv
## [1] 1
##
## $N
## [1] 2367

```

```
##
## $N_cl
## [1] 130
##
## $df
## [1] 129
##
## $nvalues
##      v2x.polyarchy.n  l.CMgnh  l.ZwvCMgwh94
## [1,]          2038      24          2283
##
## attr("class")
## [1] "ivDiag"
```

`plot_coef(g)`



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